



- 1 Title: Straw return with diverse nitrogen fertilizer application rates modulate ecosystem services
- 2 and microbial traits in a meadow soil
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# 21 Abstract:

22	Nitrogen (N) fertilization has received worldwide attention due to its benefits to soil fertility and
23	productivity, but excess N application also causes an array of ecosystem dis-services, such as
24	greenhouse gas emissions. Generally, soil microorganisms are considered to be involved in upholding a
25	variety of ecosystem services and dis-services. However, the linkages between soil ecosystem services
26	and microbial traits under different N fertilizer application rates remain uncertain. To address this, a
27	4-year in situ field experiment was conducted in a meadow soil on the Northeast China Plain after
28	straw return with the following treatments combined with regular phosphorus (P) and potassium (K)
29	fertilization: (i) regular N fertilizer (N+PK); (ii) 25% N fertilizer reduction (0.75N+PK); (iii) 50% N
30	fertilizer reduction (0.5N+PK); and (IV) no N fertilizer (PK). Ecosystem services, dis-services and
31	microbial traits responded distinctly to the different N fertilizer rates. Treatment 0.75N+PK had overall
32	positive effects on soil fertility, productivity, straw decomposition, and microbial abundance and
33	function and alleviated greenhouse effects due to N deficiency. Meanwhile, 0.75N+PK upregulated
34	aboveground biomass and soil C:N and thus increased the abundance of genes encoding
35	cellulose-degrading enzymes, which may imply the potential ability of C and N turnover. In addition,
36	most observed changes in ecosystem services and dis-services were strongly associated with microbial
37	modules and keystone taxa. Specifically, the Lasiosphaeriaceae-driven module 1 community promoted
38	straw degradation and C and N release, while the Terrimonas-driven module 3 community contributed
39	to production improvement, which was conducive to soil multifunctionality. Therefore, our results
40	suggest that straw return with 25% chemical N fertilizer reduction is optimal for achieving ecosystem
41	services. This study highlights the importance of abiotic and biotic factors in soil health and supports
42	green agricultural development by optimizing N fertilizer rates in meadow soil after straw return.





- 43
- 44 Keywords: Ecosystem services; Straw return; Nitrogen fertilization; Microbial community; Crop yield
- 45
- 46 1. Introduction

47 Multiple soil ecosystem services are indicators of soil health (Kihara et al., 2020; Lehmann et al., 48 2021). Soil ecosystem services refer to the ability of soil to function as a vital living system to sustainably increase crop productivity, improve environmental quality, tackle climate change and 49 50 promote plant and animal health (de Bello et al., 2010; Tang et al., 2019). In recent decades, 51 anthropogenic activity, such as intensive agriculture, has posed a wide range of threats to 52 agroecosystem services (Robertson et al., 2014; Allen et al., 2015). Irrational or excessive application 53 of chemical fertilizers, especially nitrogen (N), is ubiquitous to achieve high crop yields in response to 54 population surges globally (Shi et al., 2019). In fact, N is considered the essential macronutrient for all 55 biota, while excessive N fertilizer inputs not only reduce soil fertility and productivity but also lead to 56 environmental burdens (Trost et al., 2016). For example, previous studies emphasized that N fertilizer 57 abuse may accelerate greenhouse gas emissions (Huang et al., 2006; Wu et al., 2015) and degrade 58 groundwater quality (Rhymes et al., 2016). Therefore, how to achieve agroecosystem services by 59 regulating N fertilizer application rates is a critical issue that needs to be fully assessed.

Straw return has also been widely applied as a major measure to moderate soil ecosystem services (Xu et al., 2021). Plant residues, as natural organic bioenergy resources, contain abundant N that further affects soil fertility and productivity (Pan et al., 2009; Liu et al., 2014). Thus, the straw-derived N released during degradation is an important source that may serve as a partial substitute for chemical N fertilizer application (Wang et al., 2017; Latifmanesh et al., 2020). However, crop fields suffering





65	from abundant organic materials usually have low reutilization efficiency (Hou et al., 2020). Generally,
66	the majority of N in straw is released into the atmosphere as oxynitride, such as nitrous oxide (N <sub>2</sub> O),
67	resulting in lower soil organic matter (SOM) formation efficiency (Wang et al., 2019; Sun et al., 2021).
68	Subsequent literature highlighted that straw return significantly elevates greenhouse gas emissions so
69	that less than 15% of straw-derived N can be transformed into soil and become SOM (Yin et al., 2018;
70	Wu et al., 2019). However, the potential for the partial substitution of straw for chemical N fertilizer
71	application is still unclear. Revealing the mechanisms of efficient straw utilization under diverse N
72	fertilizer input rates is essential to achieving ecosystem multifunctionality.
73	Compared with plants and animals, soil contains more microorganisms living in an opaque
74	environment, making the evaluation of soil ecosystem services more complex (Handa et al., 2014;
75	Wagg et al., 2014). Agronomic management for such "multifunctionality" has prompted research into
76	the role that microbes play in providing desired rates of multiple ecosystem processes (Gong et al.,
77	2020). To our knowledge, fertilization-induced changes in microbial communities and functions are
78	fundamental to the regulation of a variety of ecosystem multifunctionalities, including SOM formation,
79	greenhouse gas emissions, litter decomposition, and crop production (Dominati et al., 2014). To date,
80	we still lack empirical evidence of the linkages among N fertilizers, specific microbial communities or
81	functions and multiple ecosystem services, and the diverse cropland services driven by complex
82	microbial traits under different N fertilizer rates are seldom clarified.
83	Microorganisms contribute to ecosystem services by modulating microbial function, community
84	composition and succession, which makes understanding the consequences of the changes in microbial
85	traits crucial for determining different N fertilizer input levels (Bradford et al., 2014). The role of

86 microorganisms in ecosystem functioning is unequivocal, and these organisms can be recognized as the





87	key drivers of ecosystem services (Chen et al., 2019a). Generally, bacteria and fungi are the main
88	drivers of straw labile and recalcitrant component decomposition, respectively (Frey et al., 2013; Ge et
89	al., 2017). Therefore, the ratio of fungi to bacteria is always considered an indicator during straw
90	degradation periods (Hogberg et al., 2007). Specifically, the expression levels of the cbhI and GH48
91	genes were identified as biomarkers of cellulolytic fungi and bacteria, respectively (Zhang et al., 2017).
92	Previous studies revealed that the N input level dominated the associations between microbial
93	composition and cellulolytic gene abundance with SOM physical fractions (Duan et al., 2021). In
94	addition, microbial module communities and keystone taxa have been used to provide satisfactory
95	explanations for ecosystem services. Chen et al. (2019b) found that particular microbial modules
96	participated in N and phosphorus (P) accumulation and CO2 emissions in a Cambisol. Moreover,
97	specific taxa are involved in agrosystem services. For example, Actinobacteria have been extensively
98	studied and can be considered the main degraders of straw by secreting cellulase (Bao et al., 2021).
99	Mortierella has been proven to increase soil fertility and crop yield due in part to its strong C
100	sequestration capacity (Ning et al., 2020). Notably, it is also well known that microbial traits are
101	mediated by nutrient availability and stoichiometry (Chen et al., 2014). C, N and P stoichiometry has
102	profound impacts on microbial in vivo metabolism and ex vivo modification processes (Chen et al.,
103	2016). Multiple studies have indicated that soil C:N and N:P ratios are the key factors mediating
104	microbial functions and soil health (Ning et al., 2020; Duan et al., 2021). Nevertheless, the knowledge
105	of the microbial mechanisms that modulate ecosystem services in response to N fertilizer input levels
106	are still rudimentary.
107	As an important grain-producing region, the Northeast China Plain contributes to more than 20%

As an important grain-producing region, the Northeast China Plain contributes to more than 20%
of the total grain yield in China (Li et al., 2017). However, excessive chemical N fertilizer inputs have





109	caused ecosystem dis-services over the past decades (Zhao et al., 2018). Therefore, a field experiment
110	was conducted to reveal the influences of N input levels on soil ecosystem multifunctionality and
111	associated microbial traits and to try to establish the linkages between them. In the present study, two
112	hypotheses were tested: (i) soil ecosystem services and dis-services would show distinct responses to N
113	fertilizer input levels, and (ii) the changes in cropland ecosystem services and dis-services would be
114	linked to specific microbial traits.
115	
116	2. Materials and methods
117	2.1 Site description and sampling
118	A field experiment under contrasting inorganic N fertilizer input levels was established in 2018 in
119	Wenchun town (44°59'61" N, 129°59'18" E), Mudanjiang city, Heilongjiang Province, Northeast China
120	Plain, which is an important grain-producing area. This region has a typical temperate continental
121	monsoon climate with an average annual temperature of 4.3 °C and a mean annual precipitation of
122	579.7 mm. The soil is classified as a meadow soil according to US Soil Taxonomy (USST). The
123	cropping system was continuous maize (Zea mays L.) monoculture. Four treatments received different
124	N fertilizer input levels after straw return to the field for 4 years as follows: (1) regular chemical
125	fertilization, N+PK (300 kg urea (N 46%) ha <sup>-1</sup> yr <sup>-1</sup> , 250 kg diammonium phosphate (P <sub>2</sub> O <sub>5</sub> 48%) ha <sup>-1</sup> yr <sup>-1</sup> ,
126	150 kg potassium chloride (K <sub>2</sub> O 50%) ha <sup>-1</sup> yr <sup>-1</sup> ); 25% reduction of N fertilizer, 0.75N+PK (225 kg urea
127	ha <sup>-1</sup> yr <sup>-1</sup> , 250 kg diammonium phosphate ha <sup>-1</sup> yr <sup>-1</sup> , 150 kg potassium chloride ha <sup>-1</sup> yr <sup>-1</sup> ); 50% reduction
128	of N fertilizer, 0.50N+PK (150 kg urea ha-1 yr-1, 250 kg diammonium phosphate ha-1 yr-1, 150 kg
129	potassium chloride ha <sup>-1</sup> yr <sup>-1</sup> ); and no N fertilizer, PK (250 kg diammonium phosphate ha <sup>-1</sup> yr <sup>-1</sup> , 150 kg
130	potassium chloride ha-1 yr-1). All straw and chemical fertilizers were applied with shallow tillage to 20





131	cm. Straw was cut into pieces less than 5 cm and input after the harvest in October, while the chemical
132	fertilizers were applied during ploughing in May of the next year. All other normal management
133	practices were consistent among treatments during the experiment. Before the experiment, the initial
134	soil contained 18.74 g kg <sup>-1</sup> SOC, 1.03 g kg <sup>-1</sup> total N and 0.54 g kg <sup>-1</sup> total P with a pH of 7.37 (H <sub>2</sub> O).
135	The yield and some of the soil chemical properties under different bulk soil treatments during the
136	experimental process are shown in Supplemental material Table S1.
137	Soils were sampled after the maize harvest in October 2021. A randomized complete block design
138	consisting of 5 treatments with 3 replications was adopted in this study. Each field plot was 4.5 m $\times$ 15
139	m. We took nine soil cores (5 cm diameter) from the top 20 cm of bulk soil in each plot. Each soil
140	sample consisted of a mixture of subsamples randomly collected at nine different positions in the same
141	plot. In total, 12 soil samples were collected from 4 treatments with 3 replicates. Soils were sieved
142	through a 2 mm mesh, the mineral particles and plant roots were carefully removed, and then the soils
143	were homogenized and stored in an incubator at 4 °C in a 40% moisture environment. One part of the
144	soil sample was air-dried to measure basal soil properties, and the other part was used for microbial
145	molecular analysis.
146	2.2 The field straw decomposition and carbon and nitrogen release experiments
147	The ditch-buried straw decomposition experiment was conducted using litter nylon bags. Maize
148	straw materials were collected after maize harvesting in 2021 and air-dried. Ten grams of maize straw
149	was cut to 2 cm in length and put into nylon litter bags, which were then sealed via heat sealing. The
150	nylon bags were 6 cm $\times$ 10 cm in size and were made of 200 mesh nylon fabric, which permitted the
1.51	

- 151 free transfer of microorganisms between the nylon bags and soil. Before maize cultivation in 2021,
- 152 litter bags containing straw were buried at 10 cm depth in a spatially random design to prevent bags





- 153 associated with a given decomposition stage being placed together in space. The litter bags were
- 154 collected after the harvest in October 2021.
- 155 The straw decomposition ratio was calculated based on dry weight loss as (dry initial mass dry
- 156 final mass)/dry initial mass. The straw-C concentration was measured by titrimetry after oxidation with
- 157 a mixture of H<sub>2</sub>SO<sub>4</sub> and K<sub>2</sub>Cr<sub>2</sub>O<sub>7</sub>. Total N, P and K were determined using the Kjeldahl, molybdenum
- 158 blue colorimetry, and flame photometry methods, respectively. All methods have been described by Lu
- 159 (2000). The initial and sampled maize straw material properties are shown in Supplemental material
- 160 Table S2. The amounts of total straw C and N released were calculated by the following equation:
- 161 The amounts of total straw C and N released = (initial C (or N) content  $\times$  dry initial mass final C
- 162 (or N) content × dry final mass) × aboveground biomass

## 163 2.3 Measurement of soil properties and assessment of ecosystem services

Soil pH was measured at a soil:water ratio of 1:2.5 (weight/weight). Air-dried soil and 25 ml of deionized water were shaken together for 1 min and left to settle for 30 min, and the soil pH was determined using an electrode. Soil organic carbon (SOC) was measured by titrimetry after oxidation with a mixture of H<sub>2</sub>SO<sub>4</sub> and K<sub>2</sub>Cr<sub>2</sub>O<sub>7</sub>. Total N and P were determined using the Kjeldahl and molybdenum blue colorimetric methods, respectively. All of these methods have been described by Lu

169 (2000).

Microbial biomass C (MBC) and microbial biomass N (MBN) were analysed using the fumigation-extraction method. Ten grams of fresh soil was fumigated with chloroform in the dark for 24 h, and then the fumigated and nonfumigated soils were extracted with 0.5 M K<sub>2</sub>SO<sub>4</sub> and shaken at 200 rpm for 0.5 h. Soil extracts were filtered through a 0.45-µm Millipore filter, and the C and N in the extracts were determined using a multi C/N 3100 analyser (Analytik Jena AG). The C and N contents





175	in extracts of the nonfumigated soil were subtracted from C and N extracted from the fumigated soil to
176	give the C and N extracted from the soil microbial biomass. Values of 0.45 and 0.54 were used to
177	calibrate the contents of MBC and MBN, respectively (Vance et al., 1987; Wu et al., 1990).
178	The activities of cellulose and N-acetyl-\beta-glucosaminidase (NAG) were measured using
179	$p$ -nitrophenyl- $\beta$ -D-cellobioside and $p$ -nitrophenyl-N-acetyl- $\beta$ -D-glucosaminide as substrates,
180	respectively. Fresh soil (1.0 g) was mixed with 2.5 mL of 0.2 M acetate buffer (pH 5.0) and 2.5 mL of
181	0.02 M substrates and then shaken at 200 rpm and 37 °C for 1 h. The reaction was stopped by adding 1
182	mL of 0.5 M CaCl <sub>2</sub> and 4 mL of 0.1 M Tris buffer (pH 12.0). The mixture was suspended with a vortex,
183	the supernatant was filtered, and the concentration of p-nitrophenol (PNP) was measured by
184	colorimetry at 400 nm. The same procedure was followed for the controls, with the exception that the
185	substrate was added after the incubation, and CaCl2 and Tris buffer were added (Dick, 2011; Geisseler
186	and Horwath, 2009).
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197	60 °C for 48 h and weighed. Aboveground biomass and crop yield were converted into weight per
198	hectare.
199	We selected 15 soil properties to estimate cropland ecosystem services, i.e., the soil fertility index
200	(SOC, total N, total P, MBC and MBN), greenhouse gas emission amount (mainly CO <sub>2</sub> , N <sub>2</sub> O and CH <sub>4</sub> ),
201	straw decomposition and C and N released, soil extracellular enzymes (cellulase and
202	N-acetyl-D-glucosaminidase), and maize biomass (aboveground biomass and crop yield). Generally,
203	SOC, total N and total P are the major soil fertility factors and indicate the present nutrient status in
204	croplands, which can be used to explain soil fertility conditions. Microbial biomass reflects ecosystem
205	productivity. Greenhouse gas emissions are related to climate change, which can be regulated by
206	fertilization regimes and soil microbial activities. Soil extracellular enzymes catalyse the
207	decomposition of a range of organic polymers, resulting in C and N turnover. Maize biomass (such as
208	aboveground biomass and crop yield) reflects soil productivity. As a whole, all of these variables
209	together contributed to the cropland function. To evaluate the function of the cropland ecosystem under
210	different fertilization conditions, we calculated an integrative soil ecosystem multifunctionality index
211	for further analysis. Due to the lack of a specific definition of multifunctionality, we first calculated the
212	Z scores of the 15 measured variables and obtained a multifunctionality value for each plot by
213	averaging the $Z$ scores of the 15 variables.
214	2.4 DNA extraction and quantification of general fungal ITS, bacterial 16S rRNA and genes
215	encoding cellulose-degrading enzymes
216	Total DNA was extracted from 0.5 g freeze-dried soil by using a Fast DNA Spin Kit for Soil
217	(MPbio, USA) according to the manufacturer's instructions and then dissolved in 50 $\mu l$ of Tris-EDTA
<b>0</b> 10	

buffer. The success of the DNA extraction was characterized by electrophoresis on 1% (wt/vol) agarose

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220	PeqLab, Germany). The extracted DNA samples were stored at $-80$ °C before molecular analysis.
221	Bacterial and fungal abundances were determined to reveal the changes in microbial community
222	compositions. The abundances of bacteria and bacteria fungi were measured according to modified
223	procedures (Fierer and Jackson., 2005). We selected the primers 338F/518R (338F:
224	CCTACGGGAGGCAGCAG; 518R: ATTACCGCGGCTGCTGG) and NSI1/58A2R (NSI1:
225	GTAGTCATATGCTTGTCT; 58A2R: CATTCCCCGTTACCCGTT) for the qPCR assay. The thermal
226	qPCR profiles for the bacteria and fungi were as follows: 95 °C 2 min, 35 cycles (95 °C 30 s, 60 °C 30
227	s, 72 °C 30 s, 80 °C 15 s), and data collection at 81 °C for 10 s; 95 °C 10 min, 40 cycles (95 °C 15 s,
228	52 °C 30 s, 72 °C 30 s, 79 °C 30 s), and data collection at 81 °C for 10 s, respectively. The initial
229	concentrations of the two plasmids used as the standards for the bacterial and fungal abundance
230	analyses were $1.22 \times 10^{10}$ and $9.05 \times 10^{9}$ , respectively.
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<ul> <li>231</li> <li>232</li> <li>233</li> <li>234</li> <li>235</li> </ul>	The fungal <i>cbhI</i> gene and bacterial <i>GH48</i> gene were selected as functional biomarkers of cellulolytic fungi and bacteria, respectively. The primers <i>GH48 F8/GH48 R5</i> ( <i>GH48_F8</i> : 5 - GCCADGHTBGGCG ACTACCT - 3; <i>GH48_R5</i> : 5 - CGCCCCABGMSWWGTACCA - 3) and <i>cbhI F/cbhI R</i> ( <i>cbhI F</i> : ACCAAYTGCTAYACIRGYAA; <i>cbhI R</i> : GCYTCCCAIATRTCCATC) were used for the qPCR assay. The abundance of bacterial <i>GH48</i> and fungal <i>cbhI</i> genes was quantified according to
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gels. The quantity and quality of DNA were checked using a Nanodrop spectrophotometer (Nanodrop,

240 plasmids as the standards for bacterial GH48 and fungal cbhI gene abundance analysis corresponded to





241	$1.85 \times 10^{\ 11}$ and $2.65 \times 10^{\ 10}$ copies g	<sup>-1</sup> dry soil, respectively. qPC	CR was performed in triplicate, and

242 amplification efficiencies higher than 95% were obtained with  $r^2$  values > 0.99.

# 243 2.5 Bacterial 16S rRNA genes and fungal ITS amplification and sequencing

244	High-throughput sequencing was performed with the Illumina MiSeq sequencing platform
245	(Illumina Inc.). Both the forward and reverse primers were tagged with an adapter and linker sequence,
246	and 8-bp barcode oligonucleotides were added to distinguish the amplicons from different soil samples.
247	The primers 515F (5'-GTGCCAGCMGCCGCGGTAA-3') and 907R
248	(5'-CCGTCAATTCMTTTRAGTTT-3') were chosen to amplify the 16S rRNA genes in the V4-V5
249	hypervariable region. PCR was conducted in a 50- $\mu L$ reaction mixture containing 27 $\mu L$ of ddH2O, 2
250	$\mu L$ (5 $\mu M)$ of each forward/reverse primer, 2.5 $\mu L$ (10 ng) of template DNA, 5 $\mu L$ (2.5 mM) of
251	deoxynucleoside triphosphates, 10 $\mu L$ of 5× Fastpfu buffer, 0.5 $\mu L$ of bovine serum albumin, and 1 $\mu L$
252	of TransStart Fastpfu polymerase (TransGen, Beijing, China). The PCR conditions were 94 °C for 5
253	min; 30 cycles of 94 °C for 30 s, 52 °C for 30 s and 72 °C for 30 s of extension; followed by 72 °C for
254	10 min (Caporaso et al., 2010).

255 The amplified primer pair ITSIF fungal ITS1 region using the was (CTTGGTCATTTAGAGGAAGTAA)/ITS2 (GCTGCGTTCTTCATCGATGC) (Ghannoum et al., 2010). 256 257 The 50- $\mu$ L reaction mixture of each reaction mix consisted of 1  $\mu$ l (30 ng) of DNA, 4  $\mu$ l (1  $\mu$ M) of each 258 forward/reverse primer, 25 µl of PCR Master Mix, and 16 µl of ddH2O. PCR amplification was 259 conducted at 98 °C for 3 min, followed by 30 cycles (98 °C for 45 s, 55 °C for 45 s, and 72 °C for 45 s), 260 with a final extension at 72 °C for 7 min (Ghannoum et al., 2010). All amplicons were cleaned and pooled in equimolar concentrations in a single tube, after which they were subjected to library 261 262 preparation, cluster generation, and 250-bp paired-end sequencing on an Illumina MiSeq platform





263 (Illumina Inc., San Diego, CA, USA).

264	The raw sequence data were processed using the Qualitative Insights into Microbial Ecology
265	(QIIME) pipeline (Caporaso et al., 2010). Sequences that fully matched the barcodes were selected and
266	distributed into separate files for the bacterial 16S rRNA and fungal ITS genes. Poor-quality sequences
267	with lengths less than 200 bp (for fungal ITS) and 500 bp (for bacterial 16S) and quality scores less
268	than 20 were discarded, and the chimaeras were removed using the UCHIME algorithm (Edgar et al.,
269	2010). The remaining sequences were assigned to operational taxonomic units (OTUs) with a 97%
270	similarity threshold using UCLUST (Edgar, 2010). Alpha diversity and Bray-Curtis distances for
271	principal coordinate analysis of the soil microbial community were calculated after rarefying all
272	samples to the same sequencing depth.
273	2.6 Statistical analysis
274	The soil ecosystem multifunctionality index, crop yields, microbial traits and other relevant soil
275	variables among treatments were subjected to a chi-square test for independence of variance.
276	Significant differences were determined by one-way analysis of variance (ANOVA) based on the post
277	hoc Tukey test at the 5% level. Prior to ANOVA, the normality and homogeneity of variances were
278	tested by the Kolmogorov-Smirnov test and Levene's test, respectively. If normality was not met, log
279	or square-root transformation was implemented. One-way ANOVA was performed using SPSS 21.0

280 (SPSS Inc., Chicago, IL, USA).

Nonmetric multidimensional scaling (NMDS) analysis was used to describe and evaluate the microbial community composition. Redundancy analysis (RDA) was performed to visualize the associations between the microbial community composition and selected soil properties. The NMDS and RDA were performed in the "Vegan" package of R (4.0.2). To describe the complex co-occurrence





285	patterns in various organisms, we constructed co-occurrence networks. We focused on the abundant
286	microbial phylotypes (with average relative abundance $> 0.01\%$ for bacteria and fungi) for network
287	construction. Nodes with Pearson correlations greater than 0.70 and $p < 0.05$ were retained. Network
288	visualization between microbial taxa was conducted by Gephi software. To obtain the keystone species
289	of each network, a Zi-Pi plot series was constructed to determine the role of each OTU. According to
290	Deng et al. (2012), the plot includes (a) peripheral nodes ( $Z \le 0.25$ , $P \le 0.62$ ). (b) module hubs ( $Z >$
291	0.25, $P \le 0.62$ ), (c) connectors ( $Z \le 0.25$ , $P > 0.62$ ) and (d) network hubs ( $Z > 0.25$ , $P > 0.62$ ). From an
292	ecological perspective, OTUs in module hubs, connectors and network hubs may be regarded as the
293	microbial keystone taxa of the network systems.
294	A heatmap was constructed to reveal the associations between microbial traits and fertilizers, soil
295	properties, greenhouse emissions and ecosystem multifunctionality. The random forest algorithm was
296	performed in the R package "RandomForest" to estimate the importance predictors of soil properties
297	and microbial traits on ecosystem multifunctionality.
298	
299	3. Results
300	3.1 Cropland ecosystem services
301	Data collection after a continuous 4-year in situ field experiment under different N input levels
302	revealed changes in cropland ecosystem services (Fig. 1). In terms of soil fertility, compared with the
303	N-limitation treatments (PK and 0.5N+PK), the SOC and total P contents were increased significantly
304	by the N+PK and 0.75N+PK treatments (Fig. 1a, c) ( $P \le 0.05$ ), while there were no significant changes
305	in the total N content (Fig. 1b). After straw decomposition (Fig. 1d), the amounts of straw C (Fig. 1e)

306 and N (Fig. 1f) released showed different responses to varying N fertilizer input levels. Generally,

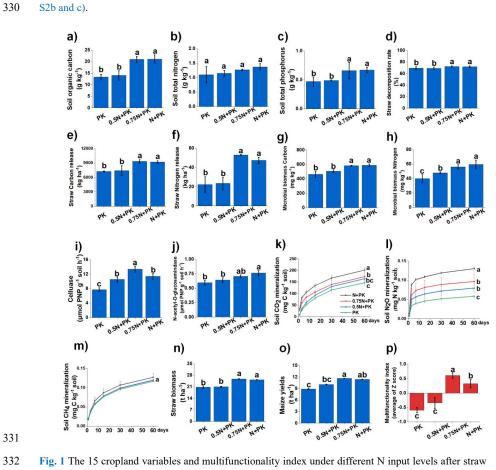




307	N-rich treatments (N+PK and 0.75N+PK) significantly increased the straw decomposition rate and
308	achieved higher amounts of straw C and N release than the N-limitation treatments ( $P < 0.05$ ).
309	However, there was no significant difference between N+PK and 0.75N+PK. Microbial biomass and
310	function were also sensitive to N fertilizer application (Fig. 1g, h, i and j). The MBC (Fig. 1g) and
311	MBN (Fig. 1h) contents were significantly higher in the N-rich treatments than in the other treatments.
312	However, the highest cellulase activity was observed in the 0.75N+PK treatment, which was
313	significantly higher than that in the other treatments (Fig. 1i) ( $P < 0.05$ ), and the
314	N-acetyl-D-glucosaminidase activity decreased with the reduction in N application (Fig. 1j).
315	For the ecosystem dis-services (greenhouse gas emissions), with the increase in N fertilizer
316	application levels, CO <sub>2</sub> and N <sub>2</sub> O emissions gradually increased (Fig. 1k, m). No significant difference
317	was observed in CH4 emissions under the different fertilization treatments (Fig. 11). In addition, the N
318	fertilizer levels also had a strong influence on maize yields and aboveground biomass (Fig. 1n, o). Our
319	results indicated that the N+PK and 0.75N+PK treatments resulted in higher maize yields and
320	above ground biomass than the other treatments ( $P < 0.05$ ), suggesting that a 25% N fertilizer reduction
321	could be satisfactory for maize growth. As expected, the 0.75N+PK treatment achieved the highest
322	multifunctionality index (0.61), followed by N+PK (0.32), 0.5N+PK (-0.34) and PK (-0.59) (Fig. 1p).
323	However, although the 0.75N+PK treatment increased the straw N release amount and may meet
324	the requirements for plant growth, the total N input was still dominated by inorganic N input (Fig. S1).
325	Therefore, the N released from the straw cannot offset the deficiency of N fertilizer. Additionally,
326	contrasting N fertilizer input levels significantly changed the stoichiometry of C, N and P (Fig. S2).
327	Notably, the 0.75N+PK treatment significantly increased the C:N ratio compared with the 0.5N+PK
328	and PK treatments ( $P < 0.05$ ). The lowest C:N ratio was shown for the 0.5N+PK treatment (Fig. S2a).







329 The N:P and C:P ratios showed no significant difference regardless of nutrient excess or limitation (Fig.

return. Abbreviations: N+PK, straw return plus regular inorganic N-P-K fertilizers; 0.75N+PK, straw
return plus regular inorganic P-K with 25% N fertilizer reduction; 0.5N+PK, straw return plus regular
inorganic P-K with 50% N fertilizer reduction; PK, straw return plus regular inorganic P-K without N
fertilizer.

# 337 3.2 Abundances of bacteria, fungi and genes encoding cellulose-degrading enzymes

338 N fertilizer input levels had marked impacts on the abundances of fungi and bacteria (Table S3).

339 The highest fungal abundance was observed in the 0.75N+PK treatment, which was significantly

16





340	higher than that in the other treatments ( $P < 0.05$ ). The N+PK treatment significantly increased
341	bacterial abundance compared with the PK treatment ( $P < 0.05$ ), while there were no obvious
342	differences among the N+PK, 0.75N+PK and PK treatments. The ratios of fungi to bacteria also
343	showed contrasting responses to N fertilization (Table. S3). The 0.75N+PK treatment significantly
344	increased the ratio of fungi to bacteria compared with the other treatments ( $P < 0.05$ ), and the lowest
345	ratio of fungi to bacteria was found in the PK treatment.

- 346 Table 1 The abundances of genes encoding ellulose-degrading enzymes
- 347

# across different N fertilizer level treatments after straw return

Treatment	<i>cbhI</i> gene abundance (×10 <sup>6</sup> copies g <sup>-1</sup> soil)	<i>GH48</i> gene abundance (×10 <sup>7</sup> copies g <sup>-1</sup> soil)	cbhI: GH48 ratio
N+PK	4.75±0.16 a	1.68±0.01 a	0.28±0.01 a
0.75N+PK	4.95±0.19 a	1.60±0.04 a	0.31±0.02 a
0.5N+PK	4.01±0.12 b	1.54±0.08 a	0.26±0.03 b
РК	3.76±0.13 b	1.40±0.06 b	0.27±0.02 b

348	The results show means $\pm$ standard deviations (n = 3). Different lowercase letters after values
349	indicate significant differences between each treatment, $P < 0.05$ . N+PK, straw return plus
350	regular inorganic N-P-K fertilizers; 0.75N+PK, straw return plus regular inorganic P-K with
351	25% N fertilizer reduction ; 0.5N+PK, straw return plus regular inorganic P-K with 50% N $$
352	fertilizer reduction; PK, straw return plus regular inorganic P-K without N fertilizer.
353	

N fertilizer input levels led to changes in the expression levels of genes encoding cellulose-degrading enzymes (Table 1). The N-rich treatments achieved higher fungal *cbhI* and bacterial *GH48* gene abundance than the N-limitation treatments. In contrast, the highest *cbhI* gene



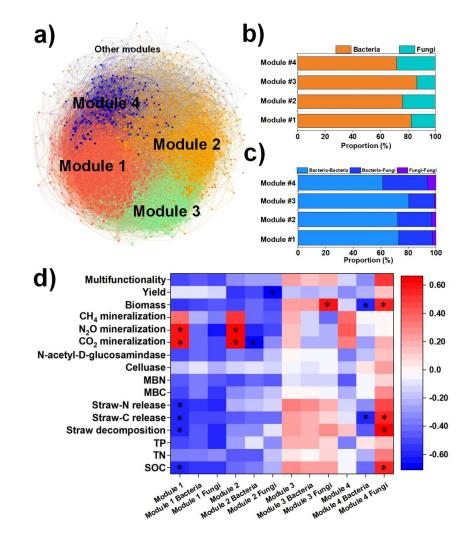


357	abundance was shown in the $0.75N+PK$ treatment, while the highest <i>GH48</i> gene abundance was shown
358	in the N+PK treatment. Compared with the PK treatment, the ratio of the fungal <i>cbhI</i> gene to the
359	bacterial <i>GH48</i> gene increased significantly under the 0.75N+PK treatment ( $P < 0.05$ ).
360	
361	3.3 Co-occurrence network analysis of the microbial community
362	Regarding fungal alpha diversities, there were no significant differences in the Chao1 index across
363	treatments. The N+PK treatment significantly increased fungal richness compared with the PK
364	treatment ( $P \le 0.05$ ) (Table S4). In addition, the PK treatment resulted in lower bacterial richness than
365	the other treatments ( $P \le 0.05$ ). No significant difference was observed in the bacterial Chao1 index
366	across treatments (Table S4). NMDS plots showed that diverse N input levels significantly changed the
367	fungal (Fig. S3a) and bacterial communities (Fig. S3b) ( $P < 0.05$ ).
368	We further conducted network analysis to identify co-occurrence patterns between specific
369	microbial taxa (Fig. 2). The cooccurrence network was aggregated into smaller coherent modules that
370	were examined to determine important module-trait relationships. The present network comprised 1963
371	nodes (composed of 1520 bacterial taxa and 443 fungal taxa) and 62206 edges with 52.49% positive
372	associations (Fig. 2a). The results showed four dominant ecological modules (1-4) that strongly
373	co-occurred within the multitrophic network, which contributed 86.10% of the whole network. Among
374	the four modules, bacteria accounted for the highest proportion in each module, contributing more than
375	70% of the total (Fig. 2b). The percentage of edges linking bacteria to bacteria (B-B) was higher than
376	that linking fungi to fungi (F-F) and bacteria to fungi (B-F). The highest proportion of B-B (80.32%)
377	was found in Module 3, while the highest proportion of B-F (32.66%) and F-F (6.00%) was found in
378	Module 4 (Fig. 2c).

379







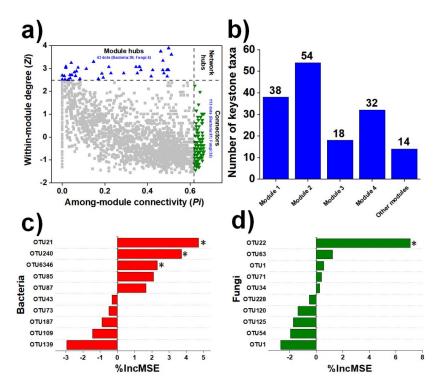
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Fig. 2 The relationships of microbial module communities with soil ecosystem services and dis-services. Multitrophic network including multiple ecological modules. The colours of the nodes represent different ecological modules (a). OTU number proportions of bacteria and fungi (b). The proportions of the edges linking bacteria to bacteria (B-B), bacteria to fungi (B-F) and fungi to fungi (F-F) in the major ecological modules (c). Links between the specific module communities with soil ecosystem services and dis-services (d). \* indicates significance at P < 0.05. Abbreviations: SOC, soil organic carbon; C: N, the ratio of the SOC content to the total N content;





- 388 N: P, the ratio of the total N content to the total P content.
- Individual nodes represented different roles in the microbial network based on the intramodule connectivity Zi and the intermodule connectivity Pi. ZP plots were constructed to identify the topological roles of each node in the network (Fig. 3a). As shown in Fig. 3b, 113 microbial taxa (81 bacterial species and 32 fungal species) were regarded as connectors, and 43 microbial taxa (39 bacterial species and 4 fungal species) were regarded as module hubs. Specifically, module 2 (54)
- 394 contained the most keystone taxa, followed by module 1 (38) and module 3 (32).



395

396 Fig. 3 The topological roles of microbial taxa and their effect on the soil multifunctionality index.

397 The topological role of each OTU was determined according to the scatter plot of within-module

398 connectivity (Z) and among-module connectivity (P) (a). The distribution of keystone taxa in each

399 ecological module (b). Contribution of bacterial (c) and fungal OTUs (d) to the soil





400 multifunctionality index. \*, \*\* and \*\*\* indicate significance at P < 0.05, 0.01 and 0.001,

## 402 **3.4** Linkage between microbial traits and soil ecosystem multifunctionality

403 The heatmap assumed close correlations between fertilizers (N input and straw), as well as soil 404 stoichiometry, and microbial traits (Fig. 4). Overall, the N input level, straw biomass and C:N ratio 405 upregulated the abundance of genes encoding cellulose-degrading enzymes. In addition, N input was 406 positively correlated with bacterial abundance, while a significant correlation was observed between 407 straw biomass and the N input level. The random forest model was also used to identify abiotic and 408 biotic attributes correlated with soil ecosystem multifunctionality (Fig. 5). The model explained 409 83.89% of the variance in ecosystem multifunctionality. The results indicated that the N input level, 410 straw biomass and soil C:N ratio were the most prominent abiotic factors affecting the ecosystem 411 multifunctionality index, while some biotic factors, such as the abundance of genes encoding 412 cellulose-degrading enzymes, significantly affected the ecosystem multifunctionality index.

413

414 Moreover, to clarify the potential main specific drivers of soil ecosystem services, correlations 415 between the microbial physiological traits and soil properties were determined to illuminate the role of 416 the microbial community in soil ecosystem multifunctionality (Fig. 3d). The results indicated that the 417 particular microbial module community was significantly correlated with soil ecosystem services. The 418 communities of modules 1 and 2 and the fungal community in module 4 showed potential in soil 419 ecosystem services (Fig. 3d). Specifically, significant correlations were observed between the SOC 420 content, straw decomposition, straw C/N release, CO2/N2O mineralization and the module 1 421 community; the module 2 community was positively correlated with greenhouse gas emissions (except

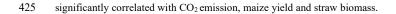
<sup>401</sup> respectively.

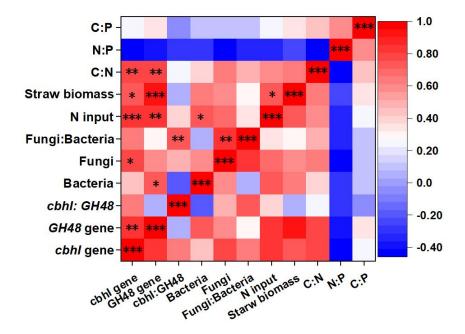




422 for CH<sub>4</sub>); and the fungal community in module 4 was positively correlated with the SOC content, straw
423 decomposition, straw C/N release and straw biomass. Furthermore, the bacterial and fungal

424 communities belonging to module 2 and the fungal community belonging to module 3 were





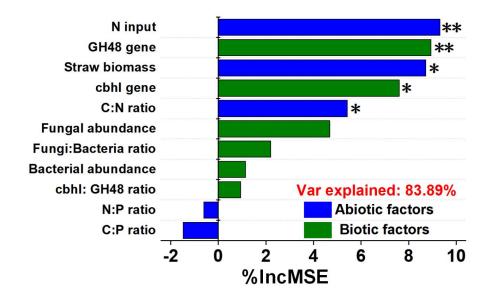
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427Fig. 4 Heatmap revealing the correlation coefficients between microbial traits with fertilization428and soil stoichiometry. \*, \*\* and \*\*\* indicate significance at P < 0.05, 0.01 and 0.001,429respectively. Abbreviations: C: N, the ratio of the SOC content to the total N content; N: P, the430ratio of the total N content to the total P content.431432432At the scale of microbial species, we selected the 20 keystone taxa (10 bacterial and 10 fungal433taxa) with the highest relative abundance for further analysis. The random forest models indicated that434the specific keystone taxa strongly influenced soil ecosystem multifunctionality (Fig. 4c and d).

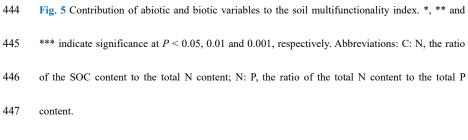




435 Bacterial OTU21 (in module 1), OTU240 (in module 2) and OTU6346 (in module 3) were highlighted 436 as essential predictors of soil ecosystem multifunctionality, and fungal OTU22 (module 3) was also 437 found to be an important variable for predicting its changes. Subsequently, the relative abundances of 438 selected keystone taxa were different across different N fertilizer level treatments after straw return 439 (Table S5). The relative abundances of fungal OTU22 and bacterial OTU21 were higher in the N-rich 440 treatments than in the N-limitation treatments. Moreover, compared with the N+PK treatment, the 441 0.75N+PK treatment increased the relative abundances of fungal OTU22 by 38.20% and bacterial 442 OTU21 by 40.63%.



443







### 449 4. Discussion

## 450 4.1 Effect of N fertilizer reduction on cropland ecosystem services after straw return

451 Soil fertility, straw decomposition, C and N release amounts, and crop productivity showed an 452 overall positive effect with the increase in the N fertilizer input level and together contributed to 453 ecosystem services. Moreover, high N applications may also cause ecosystem dis-services due to the 454 surge in greenhouse gas emissions (Fig. 1). Our results indicated that the soil fertility index (SOC, total 455 N and P contents) increased under N-rich treatments as a result of high net primary production, in 456 accordance with previous reports (Liu et al., 2010; Williams et al., 2013). Higher microbial biomass C 457 and N, as well as relevant enzyme activities, were also observed under N-rich treatments, indicating the 458 strong positive impact of abundant N fertilizer application (Fig. 1g, h, i, j). It was reported that straw 459 return with N fertilizer application can stimulate microbial activity and promote biomass accumulation 460 (Treseder, 2008). The substantially increased straw decomposition and straw C and N release under 461 N-rich treatments may be primarily attributed to the activation of microbial activity (Fig. 1d, e, f), 462 which is consistent with previous research (Ramirez et al., 2012). Recent studies have also proven that 463 rational N input can stimulate microbial ex vivo production of extracellular enzymes to accelerate 464 straw decomposition and nutrient transformation (Chen et al., 2016). Moreover, it is well known that 465 fungi have high nutrient utilization efficiency; thus, more straw-derived C and N would be stored in 466 soil under N-rich treatments than under N-limited treatments (Hou et al., 2020). Higher N availability 467 is also the premise of straw decomposition and SOM formation due to the microbial "stoichiometry 468 decomposition" theory, while the "N-mining" theory in N-limitation treatments reveals that 469 oligotrophic species (such as K-strategists) degrade native SOM because of the lack of N fertilizer 470 inputs (Chen et al., 2014). Finally, the increases in SOC, total N, and P contents and straw C and N





471	release, as well as microbial biomass and function, are commonly attributed to high aboveground
472	biomass and maize yields (Fig. 1n, o), which are favourable from the viewpoint of ecosystem services.
473	However, the overuse of N inputs also causes ecosystem dis-services, such as unintended
474	environmental consequences (Tang et al., 2019). In the present study, greenhouse gas emissions were
475	quantified to evaluate the ecosystem dis-services under different N fertilizer input levels (Fig. 1k, l, m).
476	Straw return with N fertilizer addition might be the crucial driver of CO2 and N2O emissions from
477	agroecosystems and has been widely studied in previous literature (Gregorich et al., 2005). CO2 and
478	N <sub>2</sub> O emissions increased significantly compared with those under the PK treatment, likely by
479	stimulating the activity of copiotrophs when sufficient C and N substrates were provided. For example,
480	Dieleman et al. (2010) implied that N fertilizer addition significantly increased CO2 and N2O by
481	increasing bacterial abundance through meta-analysis and field experiments, respectively. Qiu et al.
482	(2019) indicated that the emission of $CO_2$ enhanced root and mycorrhizal N uptake and increased $N_2O$
483	emissions, which was related to the changes in the soil denitrifier community composition in favour of
484	N2O-producing taxa (nirK- or nirS-type). In addition, there was no difference in CH4 emissions among
485	treatments, although contradictory results have been widely reported in previous literature (Tang et al.,
486	2019). Mapanda et al. (2011) and Liu et al. (2012) indicated that the emission of CH4 depended highly
487	on the soil water content in maize crops, which is in line with our results.
488	In summary, compared with the N+PK treatment, the 0.75N+PK treatment supported multiple
489	ecosystem services, including promoting soil fertility, straw nutrient release and microbial activity and
490	alleviating greenhouse gas emissions (Fig. 1p). Therefore, a reduction of 25% in chemical N fertilizer
491	input with straw return may be the appropriate regime to promote ecosystem services in meadow soils
492	on the Northeast China Plain.





493	4.2 Responses of the microbial composition and function to straw return with N fertilizer
494	reduction
495	Fungal and bacterial abundances, as well as the ratio of fungi to bacteria, were sensitive to the
496	changes in the N fertilizer input levels (Table S3 and Fig. 2). Straw addition with N fertilizer input
497	supplied enough C and N for microbial metabolism, thus promoting microbial proliferation (Chen et al.,
498	2016). Generally, bacterial abundance decreased with reduced N fertilizer input. This is mainly because
499	bacteria are more sensitive to N availability than fungi, which is in line with a previous study (Ramirez
500	et al., 2020). Interestingly, it is worth noting that a 25% reduction of N fertilizer significantly increased
501	fungal abundance compared with regular N inputs. This result might be attributed to the negative effect
502	of excess N fertilizer (Wan et al., 2015). Moreover, Ning et al. (2020) demonstrated that the C:N ratio
503	was the pivotal factor in fungal community compositions after performing 7 long-term field
504	experiments under different fertilization conditions across China and reported a significant positive
505	correlation between them. Therefore, the 0.75N+PK treatment with a higher C:N ratio may facilitate
506	the proliferation of microorganisms and promote an increase in microbial abundance.
507	Subsequently, our results showed that N-rich treatments resulted in higher microbial
508	cellulose-degrading gene abundances than the PK treatment (Table 1), which demonstrated the
509	irreplaceable role of N inputs in straw degradation (Zhang et al., 2017). Additionally, compared with
510	bacterial GH48 gene abundance, the increase in fungal cbhI gene abundance required adequate N
511	fertilizer inputs and was regulated by the soil C:N ratio, which suggests that rational N fertilizer inputs
512	could promote fungal function for further degradation of recalcitrant straw components (Hou et al.,
513	2020). Therefore, the ratio of <i>cbh1</i> gene abundance to <i>GH48</i> gene abundance was higher under
514	0.75N+PK than under the N-limitation treatments since the increased expression of a fungal





515	cellulose-degrading gene implies more straw C and N release.
516	Our results indicated that adequate N fertilizer upregulated fungal and <i>cbhI</i> gene abundances,
517	which may lead to multiple ecosystem services. It is therefore necessary to further explore the potential
518	associations between microbial traits and ecosystem services under diverse N fertilizer input levels.
519	4.3 Linkages of cropland ecosystem services with microbial traits
520	To clarify the effect of abiotic and biotic factors on soil ecosystem services, we then quantified the
521	contributions of abiotic and biotic attributes to the ecosystem multifunctionality index across N input
522	treatments (Fig. 4 and 5). Biotic factors, such as <i>cbhI</i> and <i>GH48</i> gene abundances, as well as abiotic
523	factors, including the C:N ratio, straw biomass and N input level, are also pivotal regulators of
524	ecosystem multifunctionality (Fig. 5). In general, promoting the rapid degradation of straw is an
525	important way to convert straw-C into SOM, thus improving soil fertility, aboveground biomass and
526	crop yield. In addition, fungi have a higher C utilization efficiency than bacteria; thus, a high fungal
527	cbhI gene abundance may achieve better soil multifunctionality (Hou et al., 2020). For abiotic factors,
528	the soil C:N ratio, straw biomass and N fertilizer input are always regarded as the main indicators of
529	soil fertility and health, likely due to providing various nutrient accessibilities and influencing the
530	microbial community composition (Ning et al., 2020).
531	Numerous studies have shown that core microbiota play a vital role in maintaining the stability of
532	soil microbial function and the complexity of microbial networks and then promoting soil nutrient
533	cycling ecosystem services (Ghannoum et al., 2015), and keystone species may show great explanatory
534	power in terms of specific network (or module) structure and function (Chen et al., 2019b). Heatmaps
535	and random forest models were used to illuminate the relationships of module communities with
536	ecosystem services (Fig. 2d and Fig 3c, d). In the present study, Terrimonas (bacterial species in





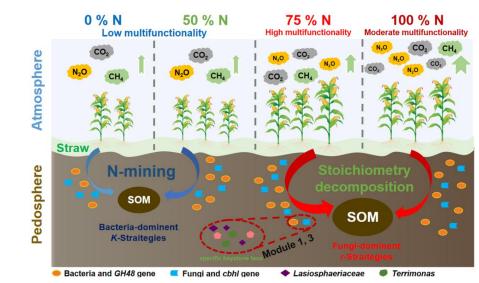
537	module 1) and Lasiosphaeriaceae (fungal species in module 3) were detected as the keystone taxa in
538	influencing soil multifunctionality of the cooccurrence network (Table S5). A previous study
539	demonstrated that straw addition significantly increased the relative abundance of Lasiosphaeriaceae,
540	which implied straw decomposition ability (Song et al., 2020). Afterwards, Lasiosphaeriaceae was
541	proven to promote straw-derived C and N accumulation by secreting multiple extracellular enzymes
542	(Guo et al., 2022). Meanwhile, Sun et al. (2023) revealed that Lasiosphaeriaceae abundance was
543	regulated by the soil C:N ratio, especially changes in mineral N. Therefore, Lasiosphaeriaceae can
544	effectively promote straw degradation and straw C and N release while driving the function and
545	community of module 1, which is consistent with our results (Fig. 2d). However, relatively few studies
546	have focused on the function of Terrimonas, so this study focused on Chitinophagaceae. As reported in
547	the previous literature, straw return was the main method to increase Chitinophagaceae abundance (Li
548	et al., 2021). Furthermore, Chitinophagaceae was indicated to have a strong ability to accumulate soil
549	C and N and degrade cellulose (Zhong et al., 2022), facilitating production improvement by regulating
550	the module 3 community and function, which is in line with our results (Fig. 2d).
551	Overall, straw return with sufficient N fertilizer application can increase the C:N ratio and
552	stimulate microbial traits, which ultimately achieve soil ecosystem multifunctionality (Fig. 6). Straw
553	return without enough N supply cannot support ecosystem services due to the decomposition of native
554	SOM and the out-of-balance microbial community composition, according to the "N-mining" theory
555	(Chen et al., 2014); straw return with sufficient N application (N+PK and 0.75N+PK) can promote soil
556	fertility, straw release, microbial activity and crop productivity, which can be explained by the

- 557 "stoichiometry decomposition" theory (Chen et al., 2014). Meanwhile, N+PK also caused more serious
- 558 ecosystem dis-services, such as greenhouse gas emissions, than the 0.75N+PK treatment. Moreover,





- 559 compared with the N+PK treatment, the 0.75N+PK treatment increased the soil C:N ratio and 560 stimulated microbial module 1 and 3 communities function, *cbh1* gene abundance, and keystone taxa
- 561 abundances, which were significantly positively correlated with soil ecosystem multifunctionality. Our
- 562 study provides evidence that a 25% reduction of chemical N fertilizer after straw return was the



563 optimal agronomic measure for ecosystem services in meadow soil on the Northeast China Plain.

565 Fig. 6 A graphical sketch of the changes in ecosystem services and potential microbial mechanisms in

# 566 response to different chemical N fertilizer application rates after straw return. N, nitrogen; SOM, soil

- 567 organic matter
- 568

564

# 569 5. Conclusion

570 Straw return combined with different chemical N fertilizer application rates significantly changed 571 ecosystem services and dis-services. Collectively, our work indicates that compared with the N+PK 572 treatment, straw return with a 25% reduction in chemical N fertilizer has the potential to improve 573 ecosystem services by maintaining soil fertility, productivity, microbial biomass and function,

574





575	0.75N+PK treatment achieved higher soil ecosystem multifunctionality than all other treatments. In
576	addition, the N input level, straw biomass and soil C:N ratio can upregulate the abundances of the <i>cbhI</i>
577	and GH48 genes, which may together achieve soil ecosystem multifunctionality.
578	Furthermore, the changes in multiple soil ecosystem services were strongly associated with
579	microbial module communities and keystone taxa. The relationships between ecosystem services and
580	microbial traits were examined here to confirm that the Lasiosphaeriaceae driving the function and
581	structure of the module 1 community leads to the promotion of straw degradation and straw C and N
582	release, while Terrimonas driving the function and structure of the module 3 community probably
583	contributes to production improvement under 0.75N+PK treatment. Therefore, a 25% reduction in
584	chemical N fertilizer with straw return might be a win-win strategy that not only produces considerable
585	ecological benefits for the pedosphere and atmosphere but also reduces fertilizer expenditures in
586	meadow soil on the Northeast China Plain.

promoting straw decomposition and C and N release and alleviating greenhouse gas emissions. The





## **Declaration of competing interests**

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

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# **Figure captions**

Fig. 1 The 15 cropland variables and multifunctionality index under different N input levels after straw return. Abbreviations: N+PK, straw return plus regular inorganic N-P-K fertilizers; 0.75N+PK, straw return plus regular inorganic P-K with 25% N fertilizer reduction; 0.5N+PK, straw return plus regular inorganic P-K with 50% N fertilizer reduction; PK, straw return plus regular inorganic P-K with 50% N fertilizer reduction; PK, straw return plus regular inorganic P-K with 50% N fertilizer reduction; PK, straw return plus regular inorganic P-K with 50% N fertilizer reduction; PK, straw return plus regular inorganic P-K without N fertilizer.

Fig. 2 The relationships of microbial module communities with soil ecosystem services and dis-services. Multitrophic network including multiple ecological modules. The colours of the nodes represent different ecological modules (a). OTU number proportions of bacteria and fungi (b). The proportions of the edges linking bacteria to bacteria (B-B), bacteria to fungi (B-F) and fungi to fungi (F-F) in the major ecological modules (c). Links between the specific module communities with soil ecosystem services and dis-services (d). \* indicates significance at P < 0.05. Abbreviations: SOC, soil organic carbon; C: N, the ratio of the SOC content to the total N content; N: P, the ratio of the total N content to the total P content.

Fig. 3 The topological roles of microbial taxa and their effect on the soil multifunctionality index. The topological role of each OTU was determined according to the scatter plot of within-module connectivity (Z) and among-module connectivity (P) (a). The distribution of keystone taxa in each ecological module (b). Contribution of bacterial (c) and fungal OTUs (d) to the soil multifunctionality index. \*, \*\* and \*\*\* indicate significance at P < 0.05, 0.01 and 0.001, respectively.

Fig. 4 Heatmap revealing the correlation coefficients between microbial traits with fertilization and soil stoichiometry. \*, \*\* and \*\*\* indicate significance at P < 0.05, 0.01 and 0.001,





respectively. Abbreviations: C: N, the ratio of the SOC content to the total N content; N: P, the

ratio of the total N content to the total P content.

Fig. 5 Contribution of abiotic and biotic variables to the soil multifunctionality index. \*, \*\* and \*\*\* indicate significance at P < 0.05, 0.01 and 0.001, respectively. Abbreviations: C: N, the ratio of the SOC content to the total N content; N: P, the ratio of the total N content to the total P content.

Fig. 6 A graphical sketch of the changes in ecosystem services and potential microbial mechanisms in response to different chemical N fertilizer application rates after straw return. N, nitrogen; SOM, soil organic matter





across different N fertilizer level treatments after straw return			
Treatment	<i>cbhI</i> gene abundance (×10 <sup>6</sup> copies g <sup>-1</sup> soil)	<i>GH48</i> gene abundance (×10 <sup>7</sup> copies g <sup>-1</sup> soil)	<i>cbhI</i> : <i>GH48</i> ratio
N+PK	4.75±0.16 a	1.68±0.01 a	0.28±0.01 a
0.75N+PK	4.95±0.19 a	1.60±0.04 a	0.31±0.02 a
0.5N+PK	4.01±0.12 b	1.54±0.08 a	0.26±0.03 b
РК	3.76±0.13 b	1.40±0.06 b	0.27±0.02 b

Table 1 The abundances of genes encoding ellulose-degrading enzymes

The results show means  $\pm$  standard deviations (n = 3). Different lowercase letters after values indicate significant differences between each treatment, P < 0.05. N+PK, straw return plus regular inorganic N-P-K fertilizers; 0.75N+PK, straw return plus regular inorganic P-K with 25% N fertilizer reduction ; 0.5N+PK, straw return plus regular inorganic P-K with 50% N fertilizer reduction; PK, straw return plus regular inorganic P-K without N fertilizer.