

1 **Microbial necromass contribution to topsoil organic carbon storage of natural**
2 **and agricultural ecosystems**

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21 **Abstract**

22 Microbial necromass is an important component of soil carbon (C). Yet, the relative
23 contribution of microbial necromass in shaping the global C stocks in agricultural and
24 natural ecosystems worldwide remains virtually unknown. Here, we compiled data on
25 fungal and bacterial necromass along with soil organic carbon (SOC) from the 0–20
26 cm soil layer across 486 study sites (145 agricultural and 341 natural ecosystems) to
27 evaluate the relative contribution of fungal necromass C (FNC) and bacterial
28 necromass C (BNC) to SOC. Our results indicated that, on average, FNC is two times
29 more important than BNC in explaining SOC in both agricultural and natural
30 ecosystems. The contributions of FNC and BNC to SOC were markedly higher in
31 agricultural ecosystems compared with natural ecosystems, with a contrasting trend in
32 the FNC/BNC ratio. Soil physicochemical properties (soil C/N ratio and clay content)
33 were the most important predictors of the contributions of FNC and BNC to SOC in
34 both ecosystems, while geographical factor (elevation) was the most important
35 predictor of the FNC/BNC ratio. Our study enhances the current level of
36 understanding regarding microbially mediated biogeochemical cycling and SOC
37 dynamics, underscoring the critical role of microbial necromass in the global C cycle.

38
39 **Keywords:** agricultural ecosystems, bacterial necromass carbon, fungal necromass
40 carbon, microbial necromass carbon, natural ecosystems

41 42 **1 Introduction**

43 As the largest carbon (C) pool in the terrestrial biosphere, soil organic carbon (SOC)
44 plays a pivotal role in shaping the global C cycle and climate system (Bellamy et al.,
45 2005; Crowther et al., 2015). In brief, plant inputs provide the primary carbon source
46 to soils, and microbial processing transforms these inputs into microbial necromass
47 that can persist over long turnover times (Angst et al., 2021; Cotrufo et al., 2013).
48 Although the living soil microbial biomass typically constitutes only about 2% of
49 SOC (a ratio referred to as the microbial quotient; Anderson & Domsch, 1989; Liu et
50 al., 2023), microbial necromass has been shown to contribute more than half and up to
51 approximately 80% of SOC, depending on soil type and analytical methods (Liang &
52 Balser, 2011; Kallenbach et al., 2016; Liang et al., 2019). In other words, microbial
53 necromass C (MNC) constitutes a substantial and critical component of stable SOC
54 (Ma et al., 2018), and its dynamics are increasingly recognized for their role in
55 regulating the terrestrial carbon cycle and climate feedbacks (Zhao et al., 2023). As
56 such, there is growing scientific attention on the forces driving the accumulation of
57 MNC and its contribution to SOC (Liang et al., 2017; Ni et al., 2020; Luo et al., 2022;
58 Zhou et al., 2023). To gain a comprehensive understanding of MNC in the global C
59 cycle, recent research has highlighted the distinct roles of fungal and bacterial
60 necromass, revealing their contrasting responses to environmental and anthropogenic
61 drivers. For instance, studies have shown that the accumulation and contribution of
62 MNC are sensitive to factors such as aridity, primary productivity, agricultural
63 management practices like tillage and fertilization, as well as key soil properties
64 including pH and clay content (Zhang et al., 2021; Zhou et al., 2023; Xu et al., 2024).

65 Despite these advances, it remains unclear whether these organism-specific
66 mechanisms translate into systematic differences in necromass contributions between
67 ecosystems under varying degrees of human interference, such as agricultural versus
68 natural systems.

69 With the distinct roles of fungi and bacteria in decomposing organic matter and
70 stabilizing organic carbon in soil, the relative contribution to SOC of fungal and
71 bacterial necromass C could be used to track the dynamics of SOC storage (Malik et
72 al., 2016). The cell walls of fungi primarily consist of chitin (a nitrogen-containing
73 polysaccharide) and β -glucans, whereas bacterial cell walls are mainly composed of
74 peptidoglycan—a complex of sugars and amino acids (Lenardon et al., 2007). As
75 bacterial amino sugars is degradable rather than fungal chitin or β -glucans, fungal
76 necromass existed in soil generally with longer turnover time than bacterial necromass
77 (Xu et al., 2022). Wang et al. (2021a) reported that the contribution of fungal
78 necromass carbon (FNC) to SOC exceeded 65%, considerably higher than that of
79 bacterial necromass carbon (BNC, 32–36%). This pattern is likely attributed to the
80 slower decomposition rate and stronger mineral-associative capacity of fungal
81 necromass. Furthermore, greater fungal biomass and higher turnover rates may
82 enhance the input flux of fungal necromass (Klink et al., 2022). The contributions of
83 FNC and BNC to SOC depended on the type of ecosystems (Wang et al., 2021a; Cao
84 et al., 2023; Xu et al., 2024). However, few studies on fungal and bacterial necromass
85 carbon and their contribution to SOC has been reported for ecosystems under human
86 interference (Zhou et al., 2023).

87 Terrestrial ecosystems can be broadly categorized into managed (agricultural) and
88 natural ecosystems (Hobbs et al., 2011; Keith et al., 2022). The agricultural
89 ecosystems are typical of plant litter derived from single crops under intensive human
90 management (Bohan et al., 2013), a context that typically leads to bacterial-dominated
91 soil communities (van Der Heijden et al., 2008). In contrast, natural ecosystems
92 display greater diversity in plant litter and root deposits (Wu et al., 2019). Within
93 these natural systems, minimal human interference enhances fungal mycelial
94 networks and stable soil aggregates, leading to higher FNC sequestration and
95 contributions to SOC (Sanaullah et al., 2020; Sae-Tun et al., 2022). While bacteria are
96 undoubtedly vital decomposers, fungi play a distinct and often dominant role in the
97 initial breakdown of complex plant polymers such as cellulose and lignin. This
98 functional prominence stems from their potent enzymatic machinery and hyphal
99 growth form, which enable physical penetration and decay of solid organic matter (de
100 Boer et al., 2005). As key decomposers, fungi are thus critical in processing cellulose
101 and other complex organic compounds (Hättenschwiler et al., 2005). Accordingly, as
102 demonstrated by Choi et al. (2018), soil cellulose-degrading genes are frequently
103 linked to fungal activity and abundance. Rather than implying higher cellulose
104 concentration per se, diverse plant inputs increase the chemical heterogeneity of
105 plant-derived polymers (e.g., cellulose, hemicelluloses, and lignin), which broadens
106 decomposer niches and often favors fungal communities in litter horizons
107 (Hättenschwiler et al., 2005; Štursová et al., 2012). In contrast, agricultural
108 monocultures tend to reduce fungal diversity unless mitigated by management

109 practices (Chen et al., 2020). Reflecting this context dependence, cellulose-rich inputs
110 can enrich saprotrophic fungi in arable soils (Clocchiatti et al., 2021), whereas
111 bacteria may contribute substantially in mineral soils or under specific microhabitat
112 and land management conditions (Štursová et al., 2012; Choi et al., 2018). Due to
113 distinct chemical properties and organo-mineral stabilization pathways, fungal and
114 bacterial necromass exhibit differing turnover times, making the FNC/BNC ratio a
115 mechanistic tracer of SOC formation (Angst et al., 2021; Kleber et al., 2021).
116 Therefore, elucidating the global distribution and drivers of FNC, BNC, and their ratio
117 across agricultural and natural ecosystems is essential for predicting
118 management-induced shifts in SOC under varying climatic and soil conditions (Zhang
119 et al., 2021; Zhou et al., 2023; Xu et al., 2024).

120 In order to explore the global patterns and drivers of FNC, BNC and the
121 FNC/BNC ratio in agricultural and natural ecosystems, we compiled data from 486
122 study sites worldwide. The aims of this study were: (1) to quantify the contributions
123 of FNC and BNC to SOC and the FNC/BNC ratio in agricultural and natural
124 ecosystems; and (2) to investigate the primary driving factors influencing the
125 contributions of FNC and BNC to SOC and the FNC/BNC ratio.

126

127 **2 Materials and methods**

128 2.1 Data collection

129 We compiled a comprehensive dataset following the stepwise workflow. (1) We
130 collected peer-reviewed papers published from 1996 to 31 December 2022 from Web
131 of Science (<http://apps.webofknowledge.com>), Google Scholar
132 (<http://scholar.google.com>), and the China National Knowledge Infrastructure
133 (<http://cnki.net>), using the keywords: ‘amino sugars’, ‘microbial necromass’,
134 ‘microbial residue’, ‘fungal residue’, and ‘bacterial residue’. Records from different
135 databases were merged and deduplicated to form an initial compilation. (2) We then
136 filtered the compiled studies to include only those focusing on topsoil, defined as the
137 0–20 cm layer. Studies reporting deeper or unspecified sampling depths (e.g., 0–30
138 cm) were excluded to ensure spatial comparability. (3) Full texts were assessed to
139 confirm the presence of paired fungal and bacterial residue data from the same
140 sample—specifically, glucosamine (GluN) and muramic acid (MurA), or directly
141 reported FNC and BNC values—to enable consistent cross-study calculation of the
142 FNC/BNC ratio. Studies lacking either biomarker were excluded from ratio analyses,
143 though those directly reporting the FNC/BNC ratio were retained. (4) Eligible
144 observations were classified into agricultural ecosystems (including dry land, irrigated
145 cropland, and submerged paddy) or natural ecosystems (forest and grassland) based
146 on study metadata. (5) For natural ecosystems, data from fertilized, polluted,
147 experimentally treated, or otherwise anthropogenically disturbed sites were excluded.
148 In total, the final dataset consisted of 2094 observations from 486 sites worldwide
149 (145 agricultural and 341 natural sites; Figure 1) reported in 164 peer-reviewed papers.
150 Of these observations, 1001 were from agricultural ecosystems, and 1093 from natural
151 ecosystems. Among the 341 natural sites, 195 were forests and 146 were grasslands.
152 For agricultural sites, we used Google Earth Engine with the LGRIP30 V1 dataset to

153 classify agricultural ecosystem into dry land and irrigated cropland, and we overlaid
154 the JRC surface-water seasonality layer to extract submerged paddy from the irrigated
155 class (LGRIP30 irrigated value = 2 and JRC seasonality ≥ 1). We ultimately classified
156 145 samples of agricultural ecosystems into 32 dry land, 72 irrigated, and 41
157 submerged paddy sites.

158 We calculated the FNC and BNC based on amino sugar concentrations following
159 widely used conversion factors, correcting total GluN for its bacterial share using
160 MurA:

$$161 \quad FNC = \left(\frac{GluN}{179.17} - 2 \times \frac{MurA}{251.23} \right) \times 179.17 \times 9 \quad \text{Equation 1}$$

162 where 9 (unitless) is the conversion factor from GluN to FNC. To estimate
163 fungal-derived GluN, we subtracted the bacterial share of GluN assuming an
164 empirical GluN:MurA molar ratio of 2:1 for bacterial residues. 179.17 and 251.23 are
165 the molecular weights of GluN and MurA, respectively. And their units are all g/mol.
166 The unit of FNC is mg/kg.

$$167 \quad BNC = MurA \times 45 \quad \text{Equation 2}$$

168 where 45 (unitless) is the conversion factor from MurA to BNC. The unit of BNC
169 is mg/kg.

170 Additional information including site geographic location (latitude and longitude),
171 topographical condition (elevation), climatic factors (MAT) and mean annual
172 precipitation [MAP]), soil physicochemical properties (pH, SOC, TN, clay content,
173 and soil temperature), and biotic (microbial and plant) factors were recorded.
174 Specifically, biotic factors included microbial biomass carbon (MBC), microbial
175 biomass nitrogen (MBN), MBC/MBN, net primary production (NPP), and
176 belowground biomass C density (BGBC). The data of topographical condition
177 (elevation) was classified as geographical factor in this study. When MAT and MAP
178 were unavailable in the original articles, we extracted them from the global climate
179 layers of WorldClim (<http://www.worldclim.org/>) with a grid precision of 30×30 arc
180 sec according to geographic location. Missing elevation data were extracted using the
181 *elevatr* package v.0.4.2 (Hollister, 2021) in the R environment. We acquired the data
182 on annual mean soil temperature from the study of Lembrechts et al. (2022), while
183 other absent soil physicochemical data were extracted from the Harmonized World
184 Soil Database
185 (<https://www.fao.org/soils-portal/data-hub/soil-maps-and-databases/harmonized-world-soil-database-v12/en/>)
186 and SoilGrids 2.0 (Poggio et al., 2021;
187 <https://www.soilgrids.org/>) using ArcGIS 10.3. We supplemented missing climatic and
188 soil variables using high-resolution, globally interpolated datasets to ensure consistent
189 spatial coverage across all sites. After retrieving missing value from gridded data, we
190 typically calibrate them against field-reported values via a field-anchored bias
191 correction (i.e., a site- or region-specific “delta” adjustment) to minimize errors
192 introduced by gridded data. While the use of such data introduces inherent
193 uncertainties, these databases are widely adopted in global-scale ecological analyses
194 and provide the most feasible approach for a unified assessment. In addition, the data
195 on NPP and BGBC were acquired from the studies of Zhao and Running (2010) and

196 Spawn et al. (2020), respectively. Missing MBC and MBN data were acquired using a
197 global database with a high resolution of 30×30 arc sec (Wang et al., 2022).

198

199 2.2 Statistical analysis

200 All the statistical analyses were performed using R v4.1.3 (R Core Team, 2021).
201 Initially, the Shapiro–Wilk test was employed to assess the normality of our data,
202 followed by the application of Levene's test to evaluate the homogeneity of variances
203 across different groups. To detect the significant differences in the contributions of
204 FNC and BNC to SOC, and the FNC/BNC ratio between agricultural and natural
205 ecosystems, as well as between forest and grassland ecosystems, the Wilcoxon rank
206 sum test was conducted. Kruskal–Wallis and Dunn's post hoc tests were used to
207 assess the significant differences of the contributions of FNC and BNC to SOC, and
208 the FNC/BNC ratio among dry land, irrigated cropland, and submerged paddy. We
209 used Spearman's rank correlation coefficient to explore the connections between the
210 16 variables considered in this study, including geographical and climatic factors, soil
211 physicochemical properties, and biotic factors. Since there was a strong positive
212 correlation between MAT and soil temperature (Figure S1), soil temperature was
213 excluded from our subsequent analyses. Linear regressions between different factors
214 and the contributions of FNC and BNC to SOC and the FNC/BNC ratio were
215 performed. Dots and smoothing curves were drawn using the *geom_point* and
216 *geom_smooth* functions, respectively, in the *ggplot2* package v.3.4.0 (Wickham,
217 2016).

218 Variation partitioning analysis was conducted using the *vegan* package v.2.5.7
219 (Oksanen et al., 2020) to evaluate the effects of four types of factors on the
220 contributions of FNC and BNC to SOC and the FNC/BNC ratio in agricultural and
221 natural ecosystems at global scale. We used a variance inflation factor threshold of 3.3
222 to eliminate those variables that were strongly correlated and avoid multicollinearity
223 (Figure S2; Kock, 2015; Fanin et al., 2020). Following factor selection, boosted
224 regression trees (BRTs) were used to partition independent influences of geographical
225 (elevation) and climatic (MAT and MAP) factors, soil physicochemical properties (pH,
226 clay, C/N, and SOC), and biotic factors (NPP, BGBC, MBC, and MBC/MBN) on the
227 contributions of FNC and BNC to SOC and the FNC/BNC ratio with the *gbm* package
228 v.2.1.8.1 (Greenwell et al., 2022).

229 Utilizing the selected factors, we performed structural equation models (SEMs)
230 to quantify the effects (direct, indirect and both) of four types of factors on the
231 contributions of FNC and BNC to SOC and the FNC/BNC ratio using *lavaan* package
232 v.0.6.19 (Rosseel, 2012). According to the previously reported potential causal
233 relationships between explanatory and response variables (Wang et al., 2021a, 2021b;
234 Li et al., 2024), we established the *priori* structural equation models for agricultural
235 and natural ecosystems, respectively (Figure S3). The SEMs were fitted via maximum
236 likelihood estimation, with non-significant paths iteratively pruned through stepwise
237 exclusion, followed by model evaluation using modification indices and
238 goodness-of-fit criteria. The fit indices included degrees of freedom (df), chi-square
239 (χ^2 , $0 \leq \chi^2/\text{df} \leq 2$), comparative fit index (CFI > 0.9), and root mean square error of

240 approximation (RMSEA < 0.08), which were used to assess the adequacy of the SEM.

241 Map, box, bar, bubble, and lollipop charts were plotted with the *ggplot2* package
242 v.3.4.0 (Wickham, 2016). To enhance map visualization, the *ggnewscale* package
243 v.0.4.8 (Campitelli, 2022) was necessary alongside the *ggplot2* package v.3.4.0
244 (Wickham, 2016). Similarly, the *ggpubr* package v.0.5.0 (Kassambara, 2022) was an
245 additional necessity when creating lollipop charts.

246

247 **3 Results**

248 3.1 Fungal and bacterial necromass contribution to SOC in agricultural and natural 249 ecosystems

250 Our analysis revealed statistically significant disparities in the contributions of FNC
251 and BNC to SOC in agricultural and natural ecosystems at the global scale ($P < 0.05$;
252 Figure 2a, b). Notably, the average contributions of FNC and BNC to SOC were
253 substantially higher in agricultural ecosystems than in natural ecosystems ($P < 0.001$;
254 Figures 2a, b). For FNC, the average contribution was 34.39% in agricultural
255 ecosystems, versus 29.24% in natural ecosystems. BNC contributed an average of
256 15.65% to SOC in agricultural ecosystems, compared to 14.02% in natural
257 ecosystems (Table 1). Our results also indicated that the contributions of FNC to SOC
258 were approximately twice those of BNC in agricultural and natural ecosystems (Table
259 1).

260 The contributions of FNC and BNC to SOC indicated no significant difference
261 between dry land and irrigated cropland ($P > 0.05$), whereas both differed
262 significantly from submerged paddy ($P < 0.05$; Figure S4a, b). In detail, across dry
263 land, irrigated cropland, and submerged paddy, the mean contributions of FNC to
264 SOC were 37.77%, 35.35%, and 22.82%, respectively, whereas those of BNC were
265 17.34%, 15.95%, and 10.55% (Table 1). Moreover, there were no significant
266 differences in the contributions of FNC and BNC to SOC between forest and
267 grassland ecosystems ($P > 0.05$; Figure S5). Specifically, FNC contributed, on
268 average, 29.11% to SOC in forests and 26.75% in grasslands, while BNC contributed
269 13.48% in forests and 14.34% in grasslands (Table 1).

270

271 3.2 Ratios of fungal and bacterial necromass in agricultural and natural ecosystems

272 Our results indicated that, at the global scale, the soil FNC/BNC ratio differs
273 significantly between agricultural and natural ecosystems ($P < 0.05$; Fig. 2c), with a
274 higher ratio in natural ecosystems (3.22) than in agricultural ecosystems (2.61; Table
275 1). The FNC/BNC ratio did not differ significantly among dry land, irrigated cropland,
276 and submerged paddy ($P > 0.05$; Figure S4c), with average FNC/BNC ratios of 2.87,
277 2.51, and 2.62, respectively (Table 1). Similarly, there was no significant difference in
278 the FNC/BNC ratio between forest and grassland ecosystems ($P > 0.05$; Figure S5),
279 and the average FNC/BNC ratios for forests and grasslands were 2.80 and 3.58 (Table
280 1), respectively.

281

282 3.3 Associations of abiotic and biotic factors with microbial necromass parameters

283 Soil physicochemical factors were the most important influence on the contributions

284 of FNC and BNC to SOC across both ecosystem types (Figures 3a–d, 4a–d).
285 Specifically, they explained 16% and 17% of the variance in the contributions of FNC
286 and BNC to SOC in agricultural ecosystems, respectively (Figures 3a, c), and 20%
287 and 24% in natural ecosystems (Figures 3b, d). BRTs corroborated this pattern, with
288 soil physicochemical factors showing the highest relative influence (51% for FNC,
289 and 44% for BNC) in agricultural systems and 44% in natural systems (Figures 4a–d).
290 All BRT models were significant ($P < 0.001$), with explained variance 36–66%. While
291 soil factors dominated overall, responses to individual variables differed between
292 ecosystems. In detail, in agricultural systems, the C/N ratio ranked third for FNC after
293 clay and SOC (Figure 4a), whereas C/N was the top predictor for FNC in natural
294 systems and for BNC in both ecosystems (Figures 4b–d). Consistently, linear models
295 showed declines in the contributions of FNC and BNC with increasing C/N in both
296 ecosystems (Figures S6g, S7g). SEMs yielded convergent results, indicating both
297 direct and indirect pathways (Figures 5a–d, 6a–d). Notably, the direct and total effects
298 of soil physicochemical factors on FNC were negative in agricultural but positive in
299 natural ecosystems (Figures 5a, b, 6a, b), whereas the effects on BNC were negative
300 in both ecosystem types (Figures 5c, d, 6c, d).

301 Our results indicated that geographical factors were the most important
302 contributors to explain the FNC/BNC ratio in both agricultural and natural ecosystems,
303 accounting for 21% and 10% of the explained variance in the FNC/BNC ratio,
304 respectively (Figures 3e, f). The results of the BRTs suggested that geographical
305 factors played a similar role in explaining the FNC/BNC ratio (Figures 4e, f). In the
306 BRT models, geographical factors emerged as the primary influencers of the
307 FNC/BNC ratio in agricultural and natural ecosystems, accounting for 32% and 44%
308 of the variance in each case, respectively (Figures 4e, f). To be more specific,
309 elevation was the most significant geographical factors influencing the FNC/BNC
310 ratio in both ecosystems (Figures 4e, f). Moreover, the FNC/BNC ratio in agricultural
311 and natural ecosystems show significantly increased with an increase elevation
312 (Figure S8a). The results of SEMs also indicated that geographical factors were the
313 most influential factors for the FNC/BNC ratio in agricultural and natural ecosystems,
314 exerting both direct and indirect effects on this ratio (Figures 5e, 6e), with the
315 standardized total effect being positive (Figures 5f, 6f).

316

317 **4 Discussion**

318 4.1 Fungal necromass contributes two times more to SOC than bacterial necromass

319 Our results show that in agricultural ecosystems, FNC/SOC ranged from 0.09% to
320 97.53% (mean \pm SE: $34.39 \pm 0.67\%$), and BNC/SOC ranged from 0.81% to 65%
321 ($15.65 \pm 0.33\%$). In natural ecosystems, FNC/SOC ranged from 0.92% to 96.29%
322 ($29.24 \pm 0.51\%$), and BNC/SOC ranged from 0.25% to 89.45% ($14.02 \pm 0.36\%$)
323 (Table 1). The FNC/BNC ratio ranged from 0.02 to 12.74 (2.61 ± 0.06) in agricultural
324 ecosystems and from 0.12 to 44.24 (3.22 ± 0.11) in natural ecosystems (Table 1).
325 Despite substantial variability at the individual sample level, the mean contribution of
326 FNC was approximately twice that of BNC in both ecosystem types. Moreover, the
327 mean FNC/BNC ratio was significantly higher in natural ecosystems than in

328 agricultural ecosystems ($P < 0.05$; Figure 2). Although this general pattern has been
329 reported in previous studies (Liang et al., 2019; Wang et al., 2021a; Zhang et al., 2023;
330 Ding et al., 2024), the systematic differences in the magnitude of these contributions
331 between agricultural and natural ecosystems—and their underlying drivers—have
332 remained poorly understood. Our study not only confirms the broad pattern but also
333 elucidates these ecosystem-level disparities and their environmental determinants.
334 Consistent with our finding that the contribution of fungal necromass carbon (FNC) to
335 SOC exceeded that of bacterial necromass carbon (BNC) in both ecosystem types
336 (Table 1), the predominance of fungal necromass may be attributed to its more
337 recalcitrant cell wall composition (e.g., chitin) and slower decomposition rate (Wang
338 et al., 2021a). Our BRT and SEM analyses further identified soil clay content and C/N
339 ratio as key drivers of FNC accumulation (Figs. 4a, 5a), reinforcing the importance of
340 organo-mineral associations in the stabilization of fungal-derived carbon.

341 Additionally, our study reveals previously unreported disparities between
342 ecosystem types, for example, the contributions of both fungal and bacterial
343 necromass carbon (FNC and BNC) to SOC were significantly higher in agricultural
344 ecosystems, while the FNC/BNC ratio was substantially elevated in natural
345 ecosystems. The higher contributions of FNC and BNC to SOC in agricultural
346 ecosystems may be attributable to two main factors. First, natural ecosystems
347 typically receive larger and more heterogeneous plant-derived carbon inputs than
348 agricultural systems. These inputs expand the plant-derived SOC pool and can dilute
349 the relative contribution of microbial necromass to SOC, thereby resulting in a lower
350 perceived contribution of microbial necromass in natural ecosystems (Angst et al.,
351 2021; Kleber et al., 2021). Second, the significantly lower soil C/N ratio in
352 agricultural ecosystems (10.78) compared to natural ecosystems (27.44) reflects
353 relative nitrogen enrichment, largely resulting from anthropogenic fertilization
354 (Castellano et al., 2015; Chen et al., 2020). This nitrogen-rich environment can
355 enhance microbial carbon use efficiency and alleviate nutrient limitation, thereby
356 promoting the production and accumulation of microbial necromass (Liang et al.,
357 2017). Supporting this mechanism, we found that the contributions of both FNC and
358 BNC to SOC decreased significantly with increasing soil C/N ratio in both
359 agricultural ecosystems (FNC/SOC: $R = -0.27$, $P < 0.001$; BNC/SOC: $R = -0.29$, $P <$
360 0.001) and natural ecosystems (FNC/SOC: $R = -0.17$, $P < 0.001$; BNC/SOC: $R =$
361 -0.35 , $P < 0.001$; Figures S6g, S7g). These results further underscore that a lower soil
362 C/N ratio—often indicative of higher nitrogen availability—is a key driver of
363 microbial necromass accumulation. It should be noted that although in situ plant
364 residues in agricultural systems (e.g., cereal straw) may have high C/N ratios, the
365 overall soil C/N ratio is reduced by management practices such as mineral fertilization
366 and the incorporation of low C/N organic amendments.

367 Furthermore, nutrient-rich conditions prevalent in agricultural systems (e.g., due
368 to fertilization) often select for bacterial-dominated communities, as many bacteria
369 exhibit *r*-strategist traits that support rapid growth under high resource availability. In
370 contrast, natural ecosystems—characterized by lower nutrient availability and greater
371 resource heterogeneity—tend to favor fungal dominance, since fungi often function as

372 *K*-strategists with higher efficiency in decomposing complex organic matter under
373 resource-limited conditions (Strickland & Rousk, 2010; Yu et al., 2022). This shift in
374 microbial community composition is observed in our results, which show a
375 significantly higher FNC/BNC ratio in natural ecosystems across our global dataset
376 (Figure 2c, Table 1). A high FNC/BNC ratio signifies a fungal-dominated
377 decomposition pathway. Fungal necromass—rich in recalcitrant compounds such as
378 chitin—is more resistant to decay, and fungal hyphae play a key role in the formation
379 of stable soil aggregates that physically protect organic matter from degradation
380 (Lenardon et al., 2007). This pathway promotes the formation of stable, long-turnover
381 SOC pools essential for long-term carbon sequestration (Six et al., 2006; Lehmann et
382 al., 2020). Furthermore, fungi generally exhibit higher carbon use efficiency than
383 bacteria, meaning a larger proportion of assimilated carbon is allocated to biomass
384 production (and subsequently necromass) rather than being respired as CO₂ (Wang &
385 Kuzyakov, 2024). Thus, the fungal-driven pathway characteristic of natural
386 ecosystems represents a highly efficient conversion of plant litter into persistent soil
387 organic matter (Kallenbach et al., 2016; Malik et al., 2016). Conversely, the lower
388 FNC/BNC ratio observed in agricultural ecosystems reflects a bacterial-dominated
389 pathway, accelerated by practices such as tillage and nutrient amendments. This
390 pathway is associated with faster carbon cycling and greater carbon loss through
391 respiration. Although microbial necromass can accumulate under these
392 conditions—sometimes contributing more significantly to a reduced total SOC
393 pool—the resulting carbon is often less stabilized (Zhou et al., 2023). Therefore, the
394 FNC/BNC ratio serves not merely as a descriptive metric, but as a functional
395 biomarker that elucidates fundamental differences in the stability and persistence of
396 SOM between managed agricultural systems and natural ecosystems.

397 Notably, as major components of agricultural ecosystems, both dryland and
398 irrigated croplands exhibited significantly greater contributions of FNC and BNC to
399 SOC than submerged paddy soils, although the FNC/BNC ratio did not differ
400 significantly among these three systems (Figure S4). This pattern may reflect similar
401 aeration regimes in dryland and irrigated systems (predominantly oxygenated),
402 leading to comparable decomposition–transformation–mineral association pathways
403 and, thus, similar net contributions of fungal and bacterial residues to SOC
404 (Ghezzehei et al., 2019). By contrast, persistent or periodic flooding in paddy soils
405 induces anoxia, suppresses aerobic decomposition, and shifts metabolic pathways
406 (e.g., denitrification and methanogenesis), potentially suppressing fungal activity or
407 dominance and altering the relative accumulation and turnover of fungal and bacterial
408 necromass (Qiu et al., 2017), resulting in contributions that differ significantly
409 from—and are lower than—those in the other two systems. Flooding can suppress
410 fungi yet also enhance the joint retention of both fungal and bacterial necromass via
411 slower decomposition and mineral protection, yielding unchanged ratios but altered
412 totals or compositional pathways (Chen et al., 2021; Gao et al., 2024).

413

414 4.2 Driving factors of the change in fungal and bacterial necromass contribution to
415 SOC and their ratio

416 Deng and Liang (2022) suggested that the potential contribution of microbial
417 necromass to the SOC pool was governed by the C/N ratio. This finding was
418 confirmed by our results (Figures 4b–d). As elaborated in *Section 4.1*, high N
419 availability (i.e., low soil C/N ratio) promotes the production and accumulation of
420 microbial necromass (Wu et al., 2025). Consequently, the contributions of both FNC
421 and BNC to SOC decreased with increasing soil C/N ratio (Figures S6g, S7g). In
422 agricultural ecosystems, high soil N levels primarily result from fertilization (Chen et
423 al., 2020). In contrast, natural ecosystems experience minimal anthropogenic
424 disturbance, N often acts as the key limiting factor for microbial activity (Elser et al.,
425 2007). Under N-limited conditions, microbes (both fungi and bacteria) allocate more
426 energy and C resources to the synthesis of N-acquiring enzymes (e.g., proteases and
427 chitinases). This shift in metabolic strategy reduces the C allocated to biomass
428 synthesis, thereby diminishing the amount of C ultimately converted into microbial
429 necromass (Mooshammer et al., 2014; Liu et al., 2024). Thus, although microbial
430 community composition differs between natural and agricultural ecosystems, the
431 regulatory role of soil C/N ratio in shaping their structure and function remains
432 consistent (Han et al., 2024). In our study, soil clay content was identified as the
433 predominant factor governing the contribution of FNC to SOC in agricultural
434 ecosystems (Figure 4a), with this contribution increasing concomitantly with clay
435 content (Figure S6d). This suggests that soils with higher clay and silt contents
436 generally accumulate greater amounts of microbial residues, particularly those derived
437 from fungi, which can be attributed to the promotion of stable organo-mineral
438 complex formation by abundant fine soil particles (Six et al., 2006 and Liang et al.,
439 2017). Furthermore, although agricultural management practices often disturb soil
440 structure, they simultaneously enhance clay enrichment and aggregate formation,
441 thereby providing effective physical protection for the long-term stabilization of
442 fungal-derived C (Chen et al., 2020; Mou et al., 2021; Zhou et al., 2023).

443 On the contrary, geographical factor (elevation) was identified as the most
444 influential predictor of the FNC/BNC ratio in both agricultural and natural ecosystems
445 (Figures 4e, f, 5f, 6f), with the ratio increasing significantly with elevation (Figure
446 S8a). Increasing elevation typically leads to decreased temperature, and increased
447 precipitation (Körner, 2007), conditions favoring fungi over bacteria due to higher
448 enzymatic capabilities and resource-use efficiency of fungi under the environments
449 (Chen et al., 2020; Yu et al., 2022; Zhang et al., 2025). High elevation also results in
450 slower soil development, which can reduce the availability of soil nutrients
451 (Guerrero-Ramírez et al., 2020). This in turn increases environmental stress and
452 restricts bacterial activity, thereby favoring the accumulation and conversion of fungal
453 residues into necromass (Li et al., 2024). Our study further demonstrated that although
454 elevation had a direct effect on the FNC/BNC ratio, it also indirectly influenced the
455 ratio by modulating climatic factors, soil physicochemical properties, and biological
456 factors (Figures 5e, 6e). This may explain why elevation is always integrate other
457 environmental factor effects in the studies of MNC (Cui et al., 2023; Zhang et al.,
458 2025).

459

4.3 Limitations and uncertainties

Although the present study provides important insights on global patterns and drivers of soil microbial necromass in agricultural and natural ecosystems, we must clarify two limitations. First, the limited data available on microbial characteristics, such as microbial community composition, enzymatic activities, and the content of soil aggregates and minerals hinder exploration of the drivers of soil microbial necromass. Second, it is undeniable that our dataset is unevenly distributed, primarily concentrating on the Northern Hemisphere, with sparse or nearly no data from other regions such as Africa, South America, and Australia (Figure 1). Additionally, the natural ecosystems in this study were limited to forests and grasslands, excluding other natural habitats such as wetlands and deserts. The uneven distribution of data may reduce the universality of MNC as a key driver of soil carbon pools in global terrestrial ecosystems. Furthermore, the compiled studies employed varied methodologies regarding sampling time, depth, and laboratory protocols. While such heterogeneity is an inherent challenge in global meta-analyses, it likely introduces additional variability and may constrain the direct comparability of certain data points. Therefore, more standardized data from these important areas and biomes are clearly required, and further investigation is warranted to fill the data gaps regarding the contribution of MNC to SOC in terrestrial ecosystems.

5 Data availability

The data and R code for this manuscript are available at <https://doi.org/10.6084/m9.figshare.28827383> (Lu, 2025).

6 Conclusions

Our results indicate that, on average, fungal necromass carbon (FNC) contributes approximately twice as much to soil organic carbon (SOC) as bacterial necromass carbon (BNC) in both agricultural and natural ecosystems. The relative contributions of FNC and BNC to SOC were found to be higher in agricultural ecosystems—an effect that is mediated by differences in soil physicochemical factors. The FNC/BNC ratio was significantly higher in natural ecosystems than in agricultural ecosystems, albeit with a modest effect size, and was primarily driven by geographical factors—particularly elevation. Our findings demonstrate that, despite considerable variability among individual sampling sites, statistically significant differences exist between agricultural and natural ecosystems in the contributions of fungal and bacterial necromass carbon (FNC and BNC) to soil organic carbon (SOC), as well as in the FNC/BNC ratio, at a global scale. These results underscore a potential fundamental divergence in the pathways and mechanisms of carbon turnover and stabilization between these two broad ecosystem types. These insights provide novel evidence that ecosystem management type (agricultural versus natural) is a key determinant of the pathways through which microbial necromass contributes to the global soil organic carbon (SOC) pool. Future studies that integrate microbial community composition with necromass dynamics across a broader range of biomes will be essential to predict ecosystem-specific responses of this critical carbon pool to

504 global change.

505

506 **Author contributions**

507 JL performed the data analysis and prepared the original draft. TC and MDB
508 contributed to manuscript review and editing. WL and HS contributed to data
509 collection. YJ contributed to data analysis. ZW supervised the project and contributed
510 to the original draft.

511

512 **Competing interests**

513 The contact author has declared that none of the authors has any competing interests.

514

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529

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829 **Table 1. Summary of the contributions of fungal necromass carbon (FNC) and**
 830 **bacterial necromass carbon (BNC) to SOC and the FNC/BNC ratio in**
 831 **agricultural and natural ecosystems at the global scale investigated in this study.**

Ecosystem	FNC/SOC (%)		BNC/SOC (%)		FNC/BNC	
	Range	Mean± SE	Range	Mean± SE	Range	Mean± SE
Natural ecosystem^{&} (<i>N</i> = 341)	0.92– 96.29	29.24 ± 0.51 b*	0.25– 89.45	14.02 ± 0.36 b	0.12– 44.24	3.22 ± 0.11 a
<i>Forest</i> (<i>N</i> = 195)	0.92– 96.29	29.11 ± 0.63 A [#]	0.94– 96.47	13.48 ± 0.43 A	0.22– 11.56	2.80 ± 0.07 A
<i>Grassland</i> (<i>N</i> = 146)	0.96– 93.89	26.75 ± 0.74 A	0.25– 89.45	14.34 ± 0.60 A	0.05– 44.24	3.58 ± 0.22 A
Agricultural ecosystem[%] (<i>N</i> = 145)	0.09– 97.53	34.39 ± 0.67 a	0.81– 65.00	15.65 ± 0.33 a	0.02– 12.74	2.61 ± 0.06 b
<i>Dry land</i> (<i>N</i> = 32)	3.01– 96.81	37.77 ± 1.15 A [#]	0.81– 65.00	17.34 ± 0.65 A	0.13– 9.12	2.87 ± 0.12 A
<i>Irrigated cropland</i> (<i>N</i> = 72)	0.09– 97.25	35.35 ± 0.73 A	1.18– 62.47	15.95 ± 0.38 A	0.02– 12.74	2.51 ± 0.06 A
<i>Submerged paddy</i> (<i>N</i> = 41)	4.96– 97.53	22.82 ± 1.55 B	1.48– 30.97	10.55 ± 0.66 B	0.31– 10.40	2.62 ± 0.16 A

832 Note: *N* refers to the number of study sites;

833 [&]Natural ecosystem includes forest and grassland;

834 [%]Agricultural ecosystem includes dry land, irrigated cropland, and submerged paddy;

835 *Within the same column, values with different lowercase letters indicate a significant
 836 difference in the same variable between agricultural and natural ecosystems
 837 (Wilcoxon rank sum test; *P* < 0.05);

838 [#]Within the same column, values with different capital letters indicate a significant
 839 difference in the same variable between forests and grasslands (Wilcoxon rank sum
 840 test), as well as between dry land, irrigated cropland, and submerged paddy (Kruskal–
 841 Wallis and Dunn’s post hoc tests; *P* < 0.05).

842 **Figure legends**

843 **Figure 1. Global distribution of the sites used in this study.** Ecosystem types are
844 distinguished by distinct shapes and colors, with the numbers in parentheses
845 indicating the number of study sites for each ecosystem type.

846 **Figure 2. Comparison of the contributions of MNC to SOC and their ratio in**
847 **agricultural and natural ecosystems.** Colors indicate different ecosystems types.
848 Significance levels: *** $P < 0.001$ and * $P < 0.05$.

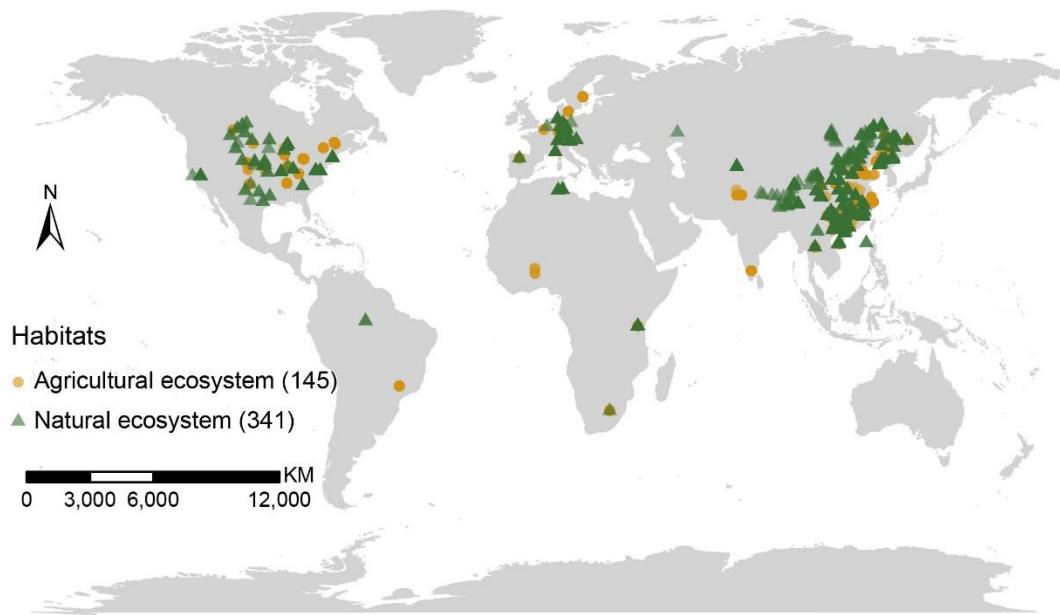
849 **Figure 3. Variations in the contributions of MNC to SOC and their ratio**
850 **explained by four types of factors in agricultural and natural ecosystems.** Colors
851 indicate different types of factors.

852 **Figure 4. Relative influence of different factors on the contributions of MNC to**
853 **SOC and their ratio in agricultural and natural ecosystems.** MAT, mean annual
854 temperature; MAP, mean annual precipitation; MBC, microbial biomass carbon; SOC,
855 soil organic carbon; C/N, the ratio of SOC to total nitrogen (TN); MBC/MBN, the
856 ratio of MBC to microbial biomass nitrogen (MBN); NPP, net primary production;
857 BGBC, belowground biomass carbon density. Colors indicate different types of
858 factors.

859 **Figure 5. The influence pathways of four types of factors on the contributions of**
860 **MNC to SOC and their ratio in agricultural ecosystems.** Direct and indirect effects
861 (a, c, e) and the standardized total effects (b, d, f) of different factors on the
862 contributions of MNC to SOC and their ratio of agricultural ecosystems are shown.
863 Standardized path coefficients representing the effect sizes of potential causal factors
864 are indicated by numbers adjacent to arrows. The width of arrows is proportional to
865 the potential causal effect between variables. The red arrows indicate positive effects,
866 and the blue arrows indicate negative effects. The numbers adjacent to boxes of
867 response variables denote the explained variance (R^2). Right-angled rectangles denote
868 single variables, whereas rounded rectangles represent composite variables. Colors
869 indicate different types of factors. Significance levels: *** $P < 0.001$, ** $P < 0.01$ and
870 * $P < 0.05$. The *priori* models are shown in Figure S3.

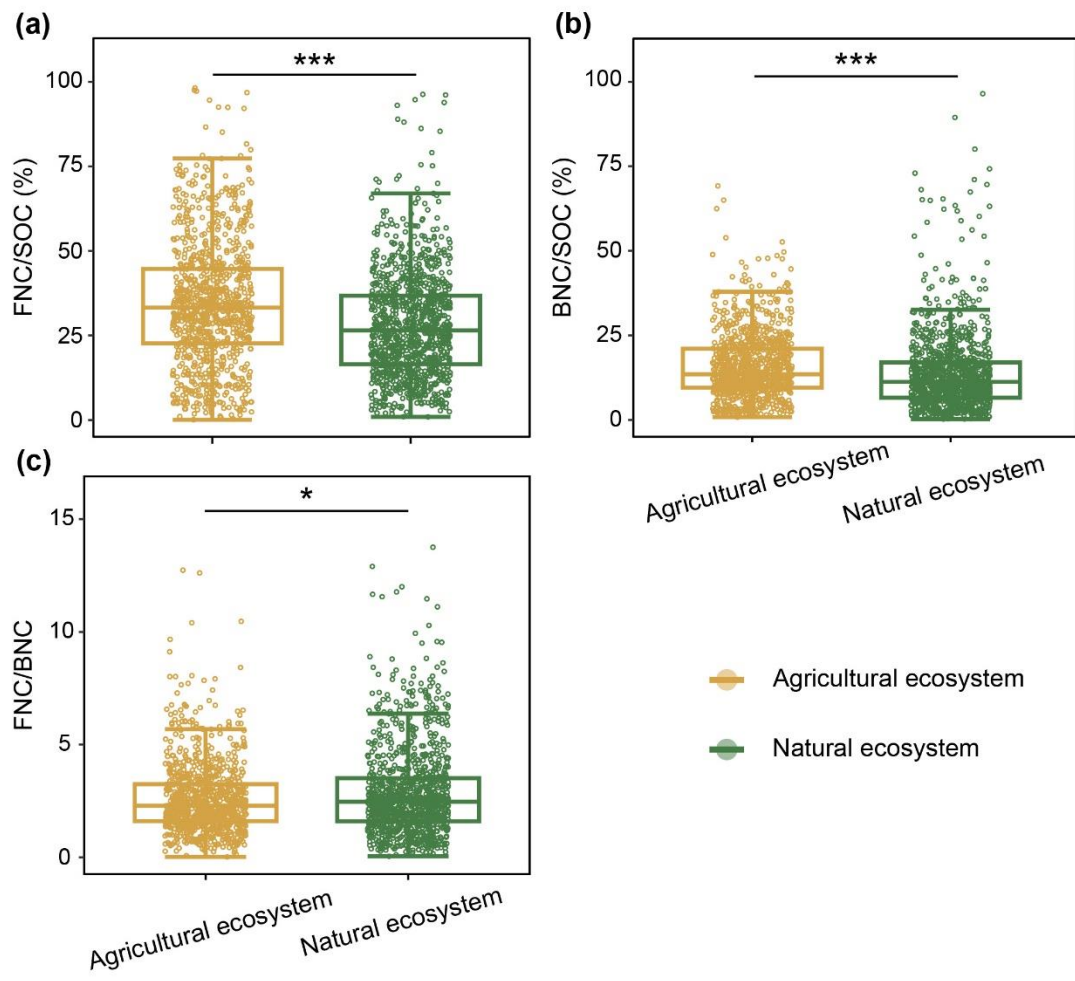
871 **Figure 6. The influence pathways of four types of factors on the contributions of**
872 **MNC to SOC and their ratio in natural ecosystems.** Direct and indirect effects (a, c,
873 e) and the standardized total effects (b, d, f) of different factors on the contributions of
874 MNC to SOC and their ratio of natural ecosystems are shown. Standardized path
875 coefficients representing the effect sizes of potential causal factors are indicated by
876 numbers adjacent to arrows. The width of arrows is proportional to the potential
877 causal effect between variables. The red arrows indicate positive effects, and the blue
878 arrows indicate negative effects. The numbers adjacent to boxes of response variables
879 denote the explained variance (R^2). Right-angled rectangles denote single variables,
880 whereas rounded rectangles represent composite variables. Colors indicate different
881 types of factors. Significance levels: *** $P < 0.001$, ** $P < 0.01$ and * $P < 0.05$. The
882 *priori* models are shown in Figure S3.

883 **Figure 1.**

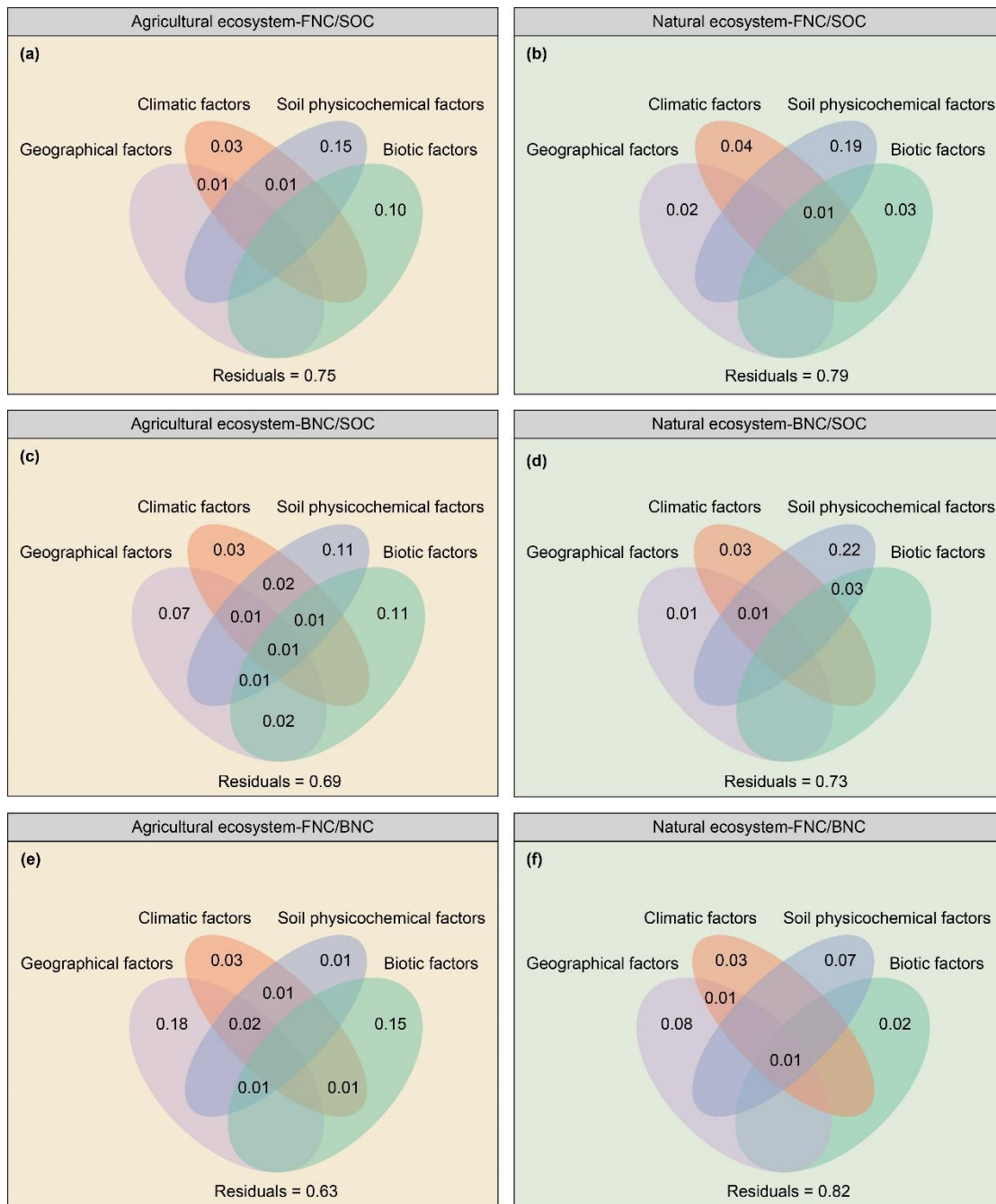


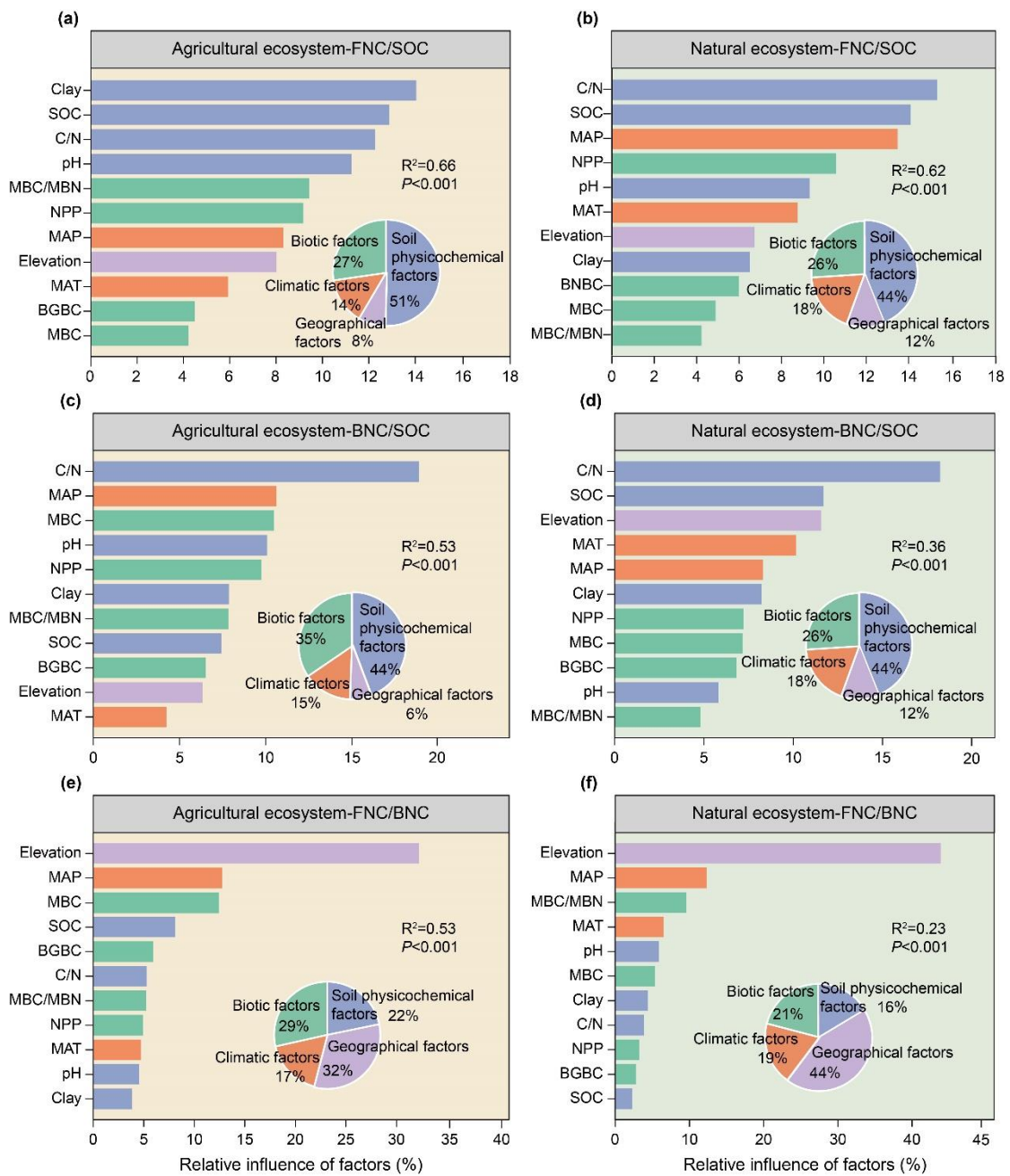
884

885 **Figure 2.**

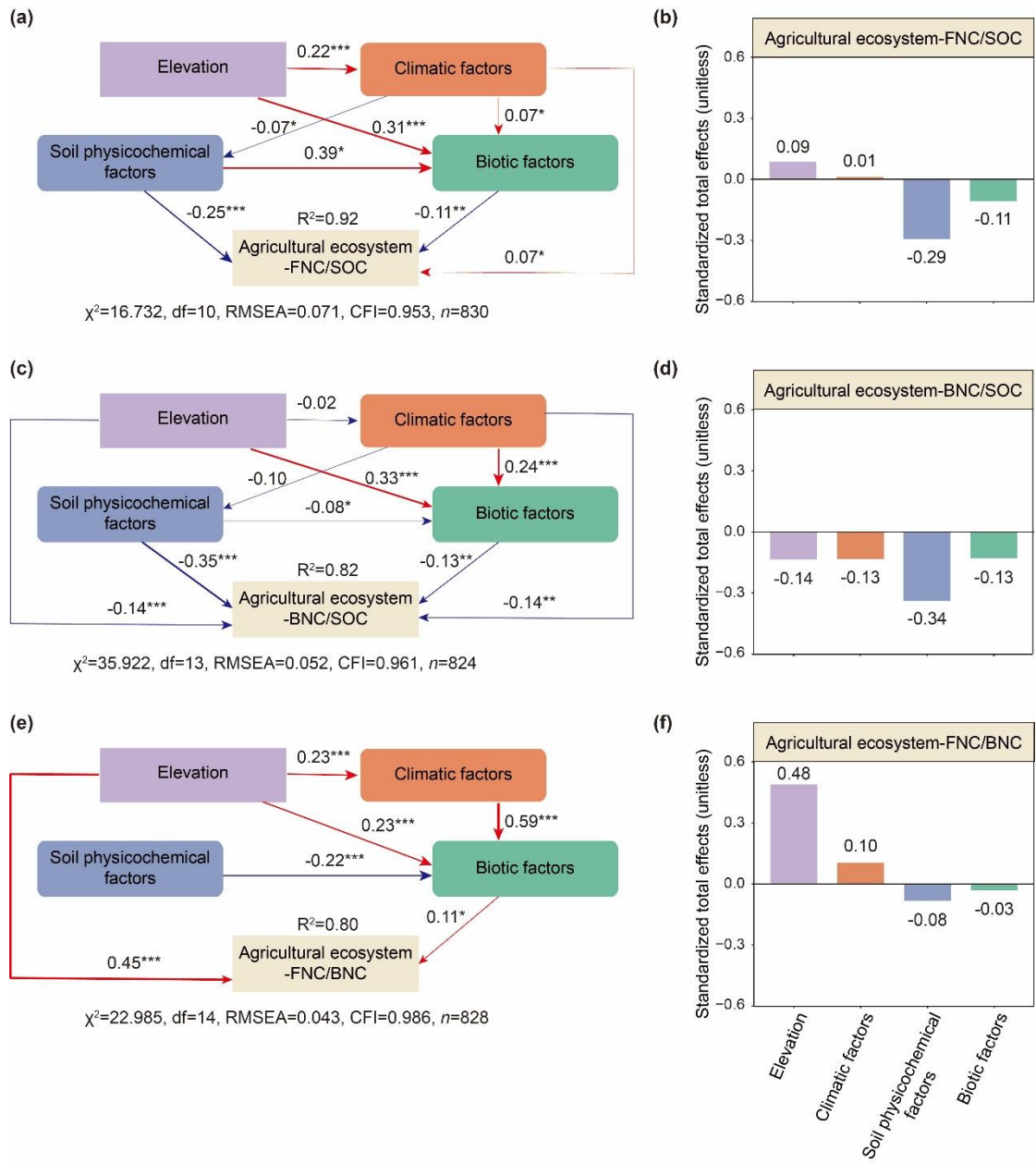


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891 **Figure 5.**



892

