1	Soil microbial biomass and community responses to experimental precipitation
2	change: a meta-analysis
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22 Abstract

23 The activity of soil microbes is strongly constrained by water availability. However, it is unclear 24 how microbial activity responds to spatial and temporal changes in precipitation, particularly to 25 long-term precipitation changes. To identify the spatiotemporal patterns of microbial responses to precipitation changes of differing durations, we conducted a meta-analysis of data from 95 26 field studies with drought treatments and 109 field studies with elevated precipitation 27 treatments. Our results indicated that microbial biomass carbon (MBC) decreased by 17% under 28 29 drought and increased by 18% under elevated precipitation. Across all studies, the phospholipid fatty acid (PLFA) biomarkers for fungi and bacteria decreased significantly under drought but 30 increased under elevated precipitation. In addition, the negative effect of drought on MBC 31 32 tended to be greater at sites with a high aridity index, but the effect of elevated precipitation on MBC did not differ among sites. More importantly, the responses of MBC, fungal and bacterial 33 PLFA abundance did not vary with treatment duration under drought, but under elevated 34 35 precipitation, they increased in the first five years of treatment and declined thereafter. These 36 results are important for our prediction of microbial responses to long-term precipitation change because they imply that microbes acclimate to long-term elevated precipitation. 37

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- Keywords: drought, precipitation change, microbial biomass, microbial community, treatment
 duration
- 42

43 **1 Introduction**

Global changes are having a substantial impact on the global hydrologic cycle via altered 44 precipitation patterns and an increased frequency of extreme wet and dry events (Bintanja and 45 46 Selten, 2014; Schlaepfer et al., 2017; Bonan and Doney, 2018). Altered precipitation regimes have a profound influence on soil processes, most notably by directly affecting soil water content 47 48 (Taylor et al., 2012; Brocca et al., 2014). Soil water content largely determines the percentage of oxygen-filled soil pores and the mobility of soil nutrients, and is therefore one of the major factors 49 mediating the biomass, activity and composition of soil microbial communities (Schimel et al., 50 2007; Brockett et al., 2012; Manzoni et al., 2012). Microbial responses to altered moisture 51 regimes are highly variable (Borken and Matzner, 2009; Balser et al., 2010) because distinct 52 53 microbial groups respond differently to variation in precipitation depending on their 54 physiological tolerances and metabolic flexibility (Allison and Martiny, 2008; Manzoni et al., 2012; Zhao et al., 2017). Fungi are generally thought to be more tolerant to water stress than bacteria 55 56 (Schimel et al., 1999; Zeglin et al., 2013), but a recent study suggests that long-term reductions 57 in precipitation can have strong impacts on soil fungal communities (Sayer et al., 2017). Clarifying the microbial responses to altered precipitation patterns is important because soil 58 59 microorganisms are key drivers of biogeochemical cycling and numerous important ecosystem processes such as decomposition and soil respiration. For example, as a result of greater fungal 60 tolerance to stress, a higher ratio of fungi to bacteria (F:B ratio) in the soil could reduce the rates 61 of organic matter turnover and soil respiration (Bailey et al., 2002; Kaisermann et al., 2015). In a 62 paddy soil, ¹³CO₂ efflux was found to increase exponentially to a maximum value with an increase 63 in the microbial biomass carbon to nitrogen ratio (C:N ratio) (Zhu et al., 2018). However, to fully 64

evaluate the impact of altered precipitation, we must also consider how differences among sites
and extant conditions influence the response of soil microbial communities to changes in soil
water availability.

68 In experimental studies, the direction and magnitude of changes in soil microbial communities in response to different precipitation regimes depend on the extant humidity levels 69 of the sites as well as the level and duration of treatments. Although gradient studies show that 70 the biomass and abundance of soil microbes increase with increasing precipitation or soil water 71 72 content (Bachar et al., 2010; Chen et al., 2015; Ma et al., 2015), the response of microbial 73 communities to change is also shaped by their acclimation to the natural variation in soil water content among sites (Evans and Wallenstein, 2012; Wallenstein and Hall, 2012). For example, 74 75 when soil microbes in drier regions are acclimated to periodic water stress, drought treatments may have little influence, whereas soil microbes acclimated to wetter regions may be more 76 sensitive to drought treatments and less responsive to elevated precipitation (Evans and 77 78 Wallenstein, 2012). However, the level of experimental drought or elevated precipitation also interacts with extant soil conditions, and the response of soil microbial communities to 79 precipitation change will also depend upon the pre-existing soil water content, because low soil 80 81 water content can induce drought-stress, whereas high soil water content reduces the oxygen concentrations in the soil. Hence, particularly high or low levels of soil moisture also represent 82 83 stressors, which can be alleviated or exacerbated by experimental treatments (Bell et al., 2014; 84 Jensen et al., 2003) and as a result, previous studies with multiple treatment levels have found that indicators of microbial community biomass, abundance, and activity consistently increased 85 at intermediate levels of supplemented rainfall (30% or 40% of mean annual precipitation; Huang 86

et al., 2015; Zhao et al., 2016). To date, most experimental studies of microbial responses to 87 altered precipitation patterns have focused on rainfall exclusion treatments (Bouskill et al., 2013), 88 and relatively fewer studies have examined the changes in microbial community composition in 89 90 response to supplemented rainfall (Brzostek et al., 2012; Cregger et al., 2012). However, 91 manipulative experiments with elevated precipitation are not only useful for studying water-92 limited systems, they can also provide critical insights into the impact of long-term global changes in precipitation regimes and patterns. Therefore, it is urgent to investigate how multiple levels of 93 94 elevated precipitation affect soil microbial community depending on site conditions and 95 experimental duration.

Last but not least, it remains unclear how soil microbes will acclimate to long-term changes 96 in precipitation regimes, which represents a significant knowledge gap for predicting ecosystem 97 responses to future climate change. Microbial communities can acclimate to new conditions 98 ('resistance') or recover more rapidly after stress ('resilience'; sensu Griffiths et al., 2000; Griffiths 99 100 and Philippot, 2013). Several long-term studies have demonstrated the increased resistance or 101 resilience of microbial communities to altered precipitation treatments, for example, in a seven-102 year experiment in a desert grassland, there was no change in microbial biomass carbon during 103 the first three years of supplemental rainfall treatments, but there were significant increases from the fourth year onwards (Bell et al., 2014). However, in a three-year study in a semiarid 104 105 grassland, there was an immediate positive effect of increased precipitation on all measured 106 microbial parameters, which increased over time (Liu et al., 2009), whereas the microbial resilience to drying-rewetting cycles was altered by 18 years of summer drought treatments in a 107 temperate heathland (Nijs et al., 2018). Given the wide range of microbial responses to altered 108

precipitation patterns and the potential influence of experimental site, treatment duration, and treatment level, it is important to identify general patterns in microbial community responses to both drought and elevated precipitation, which will help us predict changes in microbial community composition and activity.

113 Although Zhou et al. (2018) have previously presented the general patterns of soil microbe responses to precipitation changes, our meta-analysis goes further by investigating how the 114 biomass and composition of microbial communities (represented by phospholipid fatty acids, 115 116 PLFAs) responded to different levels of precipitation change. We further assessed how existing 117 drought conditions influence microbial responses to precipitation change. More importantly, we assessed the sensitivity of microbial responses to the treatment duration of precipitation change 118 119 by performing a meta-analysis of field-based manipulative experiments that reduce or elevate precipitation, respectively. We aimed to assess whether there were common patterns in the 120 responses of soil microbial communities to experimentally altered precipitation across different 121 122 climates with different treatment durations by testing the following hypotheses:

(i) Microbial biomass and community composition will respond negatively to drought
 but positively to elevated precipitation but the magnitude of the response will be
 influenced by the severity of treatments and the extant humidity levels among sites;
 (ii) Microbial physiology and community composition will acclimate to long-term
 precipitation change, and therefore, the magnitude of response of microbial biomass and
 community composition will decline with treatment duration.

129 2 Materials and methods

131 We first conducted a comprehensive search of relevant peer-reviewed articles and dissertations published from 2001 to 2018 in the Web of Science[®] and ProQuest databases using combinations 132 of the following keywords: drought, decreased precipitation, elevated precipitation, increased 133 134 precipitation, microbial, microbial biomass, microbial biomass carbon (MBC), microbial biomass 135 nitrogen (MBN), microbial community, phospholipid fatty acids (PLFAs), fungal PLFAs, bacterial 136 PLFAs, gram-positive (G+) PLFAs, and gram-negative (G-) PLFAs. We then cross-checked the 137 references of the relevant articles to identify other relevant book chapters and peer-reviewed 138 reports. We extracted mean annual precipitation (MAP, mm), mean annual temperature (MAT, °C), ecosystem type (forest, grassland, or shrubland), treatment level (%MAP), treatment 139 duration (years), soil pH, soil texture (% sand, % silt, and % clay), MBC, MBN, microbial biomass 140 141 carbon to nitrogen ratio (MBC:MBN), total PLFAs, and individual functional groups and metrics 142 defined by PLFA biomarkers in the surface (<20 cm) soil layer. When data from multiple years were given in the literature, we only used data from the most recent year to avoid temporal 143 pseudo-replication (Koricheva and Gurevich, 2014). Numerical values were extracted from 144 145 graphically presented data by digitizing the figures using Engauge Digitizer (Free Software 146 Foundation, Inc., Boston, USA). The experimental sites included in our study are shown in Figure 147 S1, which was generated by ggplot2 (Wickham, 2009). Unless otherwise stated, all data analyses were conducted using R version 3.4.2 (R Core Team, 2017) and the R packages ggsn (Baquero, 148 2017) and legendMap (Gallic, 2016). Detailed information on the meta-analysis is provided in 149 Appendix B. 150

151 2.2 Meta-analysis

The data were analysed following the methods of Hedges et al. (1999). The effect sizes for 152 drought and elevated precipitation treatments were estimated for each observation using the 153 natural log of the response ratio (RR): $\ln RR = \ln(\overline{X_t}/\overline{X_c})$, where $\overline{X_c}$ is the control mean, and 154 $\overline{X_t}$ is the treatment mean. Publication bias was estimated using a Gaussian function (Fig. S2-S3 155 in Appendix A), which was implemented in R version 3.3.3 (R Core Team, 2017); the histogram 156 of every RR for the target variables exhibited a normal distribution, indicating the absence of 157 publication bias. The average RR was calculated using the mixed model of the meta-analytical 158 software METAWIN (Sinauer Associates, Inc. Sunderland, USA). The variances of the mean effect 159 160 sizes were calculated using resampling techniques (Adams et al., 1997). If the lower boundary of the 95% confidence interval (CI) of the RR was >1, then the response was significantly positive at 161 P<0.05. If the upper boundary of the 95% CI of the RR was <1, then the response was significantly 162 negative at P<0.05. A subgroup analysis was conducted for each parameter to identify the 163 differences in the magnitude of the effects among ecosystem types. We present the results for 164 individual ecosystems only where at least three observations were available in more than one 165 subgroup. However, the results were considered valid when there were at least four observations 166 167 (Fu et al., 2011). The total heterogeneity (Q_T) was partitioned into within-group (Q_W) and 168 between-group (Q_B) heterogeneities, whereby a significant Q_B indicates a different RR among groups (Hedges et al., 1999). Group means were considered significantly different if their 95% CIs 169 did not overlap. 170

171 **2.3 Aridity index**

To compare humidity levels among sites, we used the De Martonne aridity index (AI; de Martonne, 1926), which was calculated as follows for each site:

174
$$AI = MAP/(MAT+10)$$
 (1)

where MAP is the mean annual precipitation and MAT is the mean annual temperature. A lower aridity index value corresponds to more arid conditions, whereas a higher value corresponds to a more humid climate (Liu et al., 2016).

178 **2.4 Linear and nonlinear regression analyses**

The treatment levels for altered precipitation (mm) were available in most of our data sources, 179 and we expressed them relative to the mean annual precipitation (%MAP) of the study site to 180 181 facilitate comparison. To test whether treatment level affected the sensitivity of the microbial 182 responses, we used linear regressions to examine the relationships between treatment levels 183 and the effect sizes of changes in MBC. To determine whether there was a common pattern in the response of soil microbial biomass to drought or elevated precipitation treatments among 184 185 sites with different humidity levels, we used linear regressions to examine the relationships 186 between aridity indices and the effect sizes for changes in MBC. To assess the evidence for microbial acclimation to long-term changes in precipitation, we performed linear and nonlinear 187 regressions (binomial) of the response ratios for each microbial parameter against treatment 188 duration. Finally, the relationships between MBC and microbial respiration were assessed using 189 190 linear regressions.

191 **3 Results**

3.1 Responses of soil microbial communities to altered precipitation

193 Microbial biomass and community composition were strongly modified by changes in 194 precipitation. Overall, under drought, MBC decreased by 17% (n=61) and total PLFAs decreased by 13% (n=18; Fig. 1a), whereas under elevated precipitation, MBC increased by 18% (n=56) and 195 196 total PLFAs increased by 22% (n=26; Fig. 1b). The response of MBN to drought and elevated 197 precipitation was similar to the response of MBC, and hence there was no change in the microbial 198 C:N ratio in response to either treatment. There were distinct patterns between forest and grassland ecosystems in the effects of drought, whereby a negative effect of drought on MBC 199 was observed in forests but not in grasslands (Fig. S4a). By contrast, microbial biomass and 200 201 abundance responded similarly to elevated precipitation in both ecosystem types (Fig. S4b).

The effect of drought on MBC became increasingly negative with the increased severity of the drought treatment (Fig. 2a, r^2 =0.23, P=0.0005), whereas there was no relationship between the response of MBC and the level of elevated precipitation treatments (Fig. 2b).

205 Microbial functional groups were also strongly affected by altered precipitation. Across all studies, fungal and bacterial biomarkers declined significantly with drought and increased with 206 elevated precipitation (Fig. 1a, b). Both G+ and G- bacterial biomarkers declined with drought but 207 208 were unaffected by elevated precipitation, whereas the F:B ratio increased with elevated precipitation but was unaffected by drought (Fig. 1a, b). The ratio between G+ and G- bacterial 209 210 biomarkers was not affected by either drought or elevated precipitation treatments (Fig. 1a, b). 211 The subgroup analysis of individual ecosystems revealed a strong increase in fungal biomarkers and in the F:B ratio under elevated precipitation in grasslands but not in forests (Fig. S4b). There 212 were insufficient data to evaluate the responses of G+ and G- bacteria among ecosystems. 213

However, the results showed that the responses of G- bacterial biomarkers to elevated precipitation differed significantly between acidic soils (pH<7) and alkaline soils (pH>7), with a decrease in G- bacterial biomarkers with elevated precipitation in alkaline soils (Fig. S5b).

3.2 Spatiotemporal patterns of microbial communities under simulated precipitation changes

The negative effect of drought on MBC tended to be greater at sites with a high aridity index, but the relationship was weak (Fig. 3a, r^2 =0.09, P=0.03), and the effect of elevated precipitation on MBC did not differ among sites regardless of their aridity index (Fig. 3b).

The duration of treatments included in our analysis ranged from 1 to 13 years. However, we found no evidence for an increasing impact of drought treatments over time; there was no relationship between the duration of drought treatments and the response of MBC (Fig. 4a). Accordingly, the responses of the different microbial functional groups did not change with the duration of the drought treatment. By contrast, the greatest response of MBC and fungal and bacterial PLFA abundance to elevated precipitation occurred after five years of treatment and declined thereafter (Fig. 4b, Fig. 5e, f).

3.3 Microbial respiration responses to altered precipitation

229 Microbial respiration declined with drought and increased with elevated precipitation (Fig. 6a). 230 The response of microbial respiration to elevated precipitation was strongly related to the 231 response of MBC (Fig. 6b, r^2 =0.63, P=0.011). There were insufficient studies to assess the 232 relationships between the responses of microbial respiration or MBC and drought.

233 **4 Discussion**

Our global meta-analysis revealed general patterns in the responses of microbial biomass and 234 235 community composition to experimentally modified precipitation across different ecosystems 236 and climates. We showed that drought generally led to a decrease in microbial biomass and a 237 decline in the abundance of key PLFA biomarkers, including fungal, bacterial, G+ bacterial and Gbacterial PLFAs, but did not affect microbial stoichiometry (the ratio of microbial C:N) or change 238 the ratios of key microbial functional groups. By contrast, elevated precipitation treatments led 239 to increases in the microbial biomass and the abundance of PLFA biomarkers and a higher ratio 240 241 of fungi to bacteria. Our results suggest that microbial communities can acclimate to long-term 242 elevated precipitation but not to long-term drought. In contrast to the study of Zhou et al. (2018), we analysed the responses of MBC and total PLFAs to the duration of precipitation change, which 243 is very important for the prediction of soil microbial responses to long-term precipitation changes. 244

4.1 The effect of extant site humidity levels on the response of soil microbial communities

246 The biomass of soil microorganisms is often indicated by the concentrations of MBC and MBN in the soil, both of which decreased substantially under drought treatments and increased under 247 248 elevated precipitation (Fig. 1a). This relationship not only reflects the importance of water availability for microbial growth but can also be partly attributed to variation in the availability of 249 organic substrates (Fig. S7, 8, Zhou et al., 2018; Wei et al., 2019). We also found that the 250 251 sensitivity of MBC to drought increased with the treatment level (Fig. 2a). More importantly, the 252 sensitivity of MBC to drought was greater at relatively humid sites (Fig. 3a, Fig. S10c), possibly 253 because microbial communities at drier sites are more likely to acclimate to low water availability 254 (Evans and Wallenstein, 2012). This difference in sensitivity was also reflected by the microbial responses in different ecosystems, as MBC and MBN declined significantly with drought in forests 255

but not in grasslands (Fig. S4a), possibly because the aridity index of the forest sites (~60) in our study was significantly higher than that of the grassland sites (~39; Fig. S6). However, the magnitude of the positive response of MBC to elevated precipitation was not related to treatment levels or aridity indices (Fig. 2b, Fig. 3b), which is consistent with the results of MBC responses to elevated precipitation between forest and grassland ecosystems (Fig. S4b).

We found no influence of drought on the C:N ratio of the microbial biomass (Fig. 1a), which 261 suggests that drought constrains microbial growth without altering microbial stoichiometry. This 262 263 is noteworthy because the microbial C:N ratio is generally considered a better indicator of 264 changes in soil C and N cycling (Buchkowski et al., 2015) than microbial biomass, and our results suggest that microbial communities maintain a relatively stable C:N ratio under drought 265 conditions. The C:N ratio of fungi (~15) is higher than that of bacteria (~6; Wallenstein et al., 266 2006), and given that fungi and bacteria are the dominant decomposers in soils (Manzoni et al., 267 2012; Schimel et al., 2007), the relatively stable microbial C:N ratio could also be related to the 268 269 ratio of fungi to bacteria (F:B) in the soil, which did not change under drought (Fig. 1a). However, the F:B ratio increased under elevated precipitation without a concomitant change in the 270 271 microbial C:N ratio (Fig. 1b). Hence, the mechanisms underlying the stable microbial C:N 272 stoichiometry require further investigation.

Fungi and bacteria have distinct physiologies that underlie their responses to environmental stressors (Manzoni et al., 2012; Schimel et al., 2007). Fungi are generally considered droughttolerant (Schimel et al., 2007; Strickland and Rousk, 2010; Manzoni et al., 2012), whereas bacteria may be more drought-sensitive (Holland and Coleman, 1987; Manzoni et al., 2012), and we therefore expected to see a greater decline in bacterial PLFAs under drought. However, both

fungal and bacterial PLFAs declined significantly in drought treatments, and their responses were 278 279 sufficiently similar that the F:B ratio was not altered by drought (Fig. 1a). For example, a long-280 term manipulative precipitation experiment also showed that decreased precipitation did not 281 change the F:B ratio, although the original drier plots had higher F:B ratios than the original 282 wetter plots (Zeglin et al., 2013). The absence of change in the F:B ratio suggests that the recalcitrance of soil C may not be affected by drought, as soil C is expected to be more persistent 283 when its storage is mediated by the fungal biomass and to be more labile when mediated by the 284 285 bacterial biomass (Ho et al., 2017). Conversely, the increases in fungal and bacterial PLFA 286 abundance as well as the F:B ratio under elevated precipitation (Fig. 1b) are possibly related to greater availability of soil C sources as suggested by the increase in dissolved organic carbon (DOC) 287 under elevated precipitation (Fig. S7b). This is consistent with the results from Ge et al. (2017a), 288 which found that shoot and root biomass increased with N fertilization as did fungal and bacterial 289 biomarkers and F:B ratios. 290

291 Different groups of bacteria also have distinct physiologies and life cycles that confer varying 292 levels of stress-tolerance. Gram-positive bacteria, which have a thicker peptidoglycan layer in 293 their cell walls, are generally considered to be more resistant to drought than gram-negative 294 bacteria (Lennon et al., 2012; Schimel et al., 2007; Manzoni et al., 2012). However, across all studies, both gram-positive and gram-negative bacterial PLFAs declined similarly under drought 295 296 and did not change with elevated precipitation (Fig. 1a). It is possible that the low number of 297 studies reporting data on individual taxonomic groups precludes the detection of a clear pattern 298 across multiple ecosystems and climates (Fig. S4). Alternatively, the timing of measurements and the availability of substrates may play a more important role than the experimental treatment 299

per se; Gram-negative bacteria predominantly use labile plant-derived C as a substrate, whereas 300 301 Gram-positive bacteria can use more recalcitrant C sources derived from soil organic matter 302 (Griffiths et al., 1999; Fierer et al., 2003; Kramer and Gleixner, 2006; Potthoff et al., 2006), and 303 hence their resistance and resilience to drought treatments are also likely to differ as a result of 304 changes in available resources. Although Gram-negative bacteria predominantly use labile plant-305 derived C as a substrate, they tend to favour acidic soils (Ge et al., 2017a), which explains why Gram-negative bacteria did not increase under elevated precipitation or with increased DOC 306 307 concentration at higher soil pH (Fig. S7b). Indeed, when we calculated the effect sizes for soils 308 with pH<7 (acidic soils) or pH>7 (alkaline soils), we found that the responses of G- bacterial biomarkers to elevated precipitation differed significantly between acidic soils (pH<7) and 309 310 alkaline soils (pH>7), with a decrease in G- bacterial biomarkers in alkaline soils (Fig. S5b).

4.2 Evidence for microbial acclimation to long-term elevated precipitation but not to long-term drought treatments

313 Our meta-analysis demonstrated changes in the sensitivity of multiple microbial parameters to 314 altered precipitation in field experiments lasting more than 10 years. We expected diminishing 315 impacts of the treatments with increasing experimental duration as the microbial communities acclimated to altered precipitation patterns. However, we found no relationship between 316 317 microbial responses to drought treatments and experimental duration (Fig. 4a), possibly because the majority of drought experiments had a duration of <5 years. However, it is noteworthy that 318 319 the response of MBC was lower in the few drought experiments with a duration >5 years (Fig. 320 4a). By contrast, the responses of MBC and fungal and bacterial PLFAs declined with the increasing duration of elevated precipitation treatments (Fig. 4b, Fig. 5e, f). Individual studies 321

have observed changes in microbial biomass and community structure during continuous multi-322 323 year (5-6 years) treatments with increased precipitation and found that treatment duration 324 affected the magnitude of the microbial response to elevated precipitation (Bell et al. 2014; 325 Gutknecht et al. 2012). Together, these results suggest that drought stress has a large and 326 immediate impact on soil microbial communities, but that microbial communities are better at 327 acclimating to elevated precipitation over time. Several mechanisms could underlie the acclimation of microbial communities to increased precipitation, including the acclimation of 328 329 individual microbial taxa (Allison et al., 2010; Crowther and Bradford, 2013), shifts in microbial 330 community composition (Barcenas-Moreno et al., 2009; Treseder et al., 2016), and evolutionary adaptation of microbial populations (Romero-Olivares et al., 2015). These mechanisms are not 331 332 mutually exclusive, and their relative influence may vary with season (Contosta et al., 2015), ecosystem, and time scale. The observed shifts in the fungal to bacterial ratio (Fig. 1b) could 333 facilitate the acclimation of microbial communities to increased precipitation across sites, but 334 335 further work is required to determine the mechanisms and the pace of microbial acclimation to climate changes to improve predictions of the long-term consequences of altered precipitation 336 patterns for key soil microbial processes. Importantly, our results provide some evidence that the 337 338 long-term impact of altered precipitation on soil microbial communities may be weaker than suggested by initial or short-term responses, and long-term experiments are therefore crucial for 339 340 predicting future change.

The responses of decomposition processes to climate change depend on microbial community composition (Glassman et al., 2018), and microbial respiration during decomposition is an important component of soil CO₂ efflux. Changes in precipitation can influence microbial

respiration via the impact on microbial biomass and community composition (Huang et al., 2015; 344 345 Ren et al., 2018). Our results showed that the extent of the changes in microbial respiration under 346 elevated precipitation (Fig. 6a) was strongly related to the sensitivity of the microbial biomass 347 (Fig. 6b). Although there are currently insufficient data to predict whether microbial respiration will acclimate to long-term variation in precipitation, the relationship between MBC and 348 microbial respiration suggests that the response in respiration will most likely parallel the 349 observed decline in the response of MBC after 5 years of elevated precipitation (Fig. 6b). The lack 350 351 of data on microbial activity in response to drought represents a significant knowledge gap in our 352 understanding of the impact of altered precipitation patterns on ecosystem functioning.

353 **5 Conclusions**

Given the importance of microbial processes in soil functioning and plant productivity, our meta-354 355 analysis informs future global change research by identifying the spatiotemporal patterns in the 356 response of microbial biomass and community composition to altered precipitation regimes. Our results suggest that microbial communities can acclimate to long-term elevated precipitation but 357 not to long-term drought. In contrast to the study by Zhou et al. (2018), we analysed the 358 359 responses of MBC and PLFAs to experimental duration, which is important for predicting the 360 responses of soil microbes to long-term precipitation changes. However, we still lack long-term 361 experiments (>10 years duration) that focus on soil microbial activity and function, particularly in 362 response to drought. Detailed studies on soil microbial community composition, specific functional groups and key microbial processes are required to improve predictions of future 363 ecosystem functioning under altered precipitation regimes. 364

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374 Author contributions

375 SX and CL designed this study. SX conducted the literature searches, and SX and WXG collected 376 the data. SX analysed the data and wrote the paper. EJS, GZ and PZ contributed to data 377 interpretation and commented on the details of the manuscript drafts.

378 **Conflict of Interest**

379 The authors declare that they have no conflict of interest.

380 Data availability

381 The data set of the peer-reviewed publication will be accessible after acceptance.

382

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Fig. 1 Responses of microbial biomass and microbial communities to altered precipitation, i.e., (a) drought and (b) elevated precipitation; numbers in parentheses represent the number of observations for each parameter, and error bars represent 95% confidence intervals. Error bars overlapping the dashed vertical line indicate no effect relative to the controls, where MBC is microbial biomass carbon, MBN is microbial biomass nitrogen, PLFAs is phospholipid fatty acids, F:B is the ratio of fungal to bacterial PLFAs, G+ is gram-positive bacteria, G- is gram-negative bacteria, G+:G- is the ratio of gram-positive bacterial to gram-negative bacterial PLFAs.



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563 Fig. 2 Relationships between the level of drought or elevated precipitation treatments, expressed 564 relative to the mean annual precipitation (%MAP), and the response ratio of MBC under (a) 565 drought and (b) elevated precipitation, where RR is response ratio and other abbreviations follow 566 567 the legend for Figure 1. The symbol size (n) represents the number of replicates in each study 568 and the grey shading represents the confidence interval around the mean (line) when the relationship is significant at P<0.05. 569

Increased precipitation





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Fig. 3 Relationships between the aridity indices and the response ratio of MBC under (a) drought and under (b) elevated precipitation. The abbreviations follow the legend for Figure 2. The symbol size (n) represents the number of replicates in each study and the grey shading represents the confidence interval around the mean (line) when the relationship is significant at *P*<0.05.



Drought

Increased precipitation



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Fig. 4 Relationships between treatment duration (years) and the response ratio of MBC under (a) drought and under (b) elevated precipitation. The abbreviation yr indicates year; other abbreviations follow the legend for Figure 2. The symbol size (n) represents the number of replicates in each study and the grey shading represents the confidence interval around the mean (line) when the relationship is significant at *P*<0.05.



Increased precipitation



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Fig. 5 Relationships between treatment duration in years (yr) and the response ratios (RR) of microbial functional groups for (a) fungal PLFAs, (b) bacterial PLFAs, (c) G+ PLFAs, (d) G- PLFAs under drought, and for (e) fungal PLFAs, (f) bacterial PLFAs, (g) G+ PLFAs, (h) G- PLFAs under elevated precipitation; PLFAs are phospholipid fatty acids, G+ is gram-positive bacteria and G- is gram-negative bacteria. The symbol size (n) represents the number of replicates in each study and the grey shading represents the confidence interval around the mean (line) when the relationship is significant at *P*<0.05.

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Fig. 6 The response of microbial respiration under drought and elevated precipitation (a), and the relationship between microbial respiration and MBC (b). MR is microbial respiration; other abbreviations follow the legend for Figure 1. The vertical line is drawn at an effect size of one (i.e., no effect), and the error bars that overlap the dashed line are not significant. The symbol size (n) represents the number of replicates in each study and the grey shading represents the confidence interval around the mean (line) when the relationship is significant at P<0.05.

