

1 **Mixing crop residues induces a synergistic effect on microbial biomass and an additive**
2 **effect on soil organic matter priming**

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8 **Abstract**

9 Applying crop residues is a widely used strategy to increase soil organic matter (SOM) in
10 arable soils because of its recorded effects on increasing microbial biomass and
11 consequently necromass. However, fresh residue inputs could also “prime” the
12 decomposition of native SOM, resulting in accelerated SOM depletion and greenhouse gas
13 (GHG) emission. Increasing the botanical diversity of the crops grown in arable systems has
14 been promoted to increase the delivery of multiple ecological functions, including increasing
15 soil microbial biomass and SOM. Whether mixtures of fresh residues from different crops
16 grown in polyculture contribute to soil carbon (C) pools to a greater extent than would be
17 expected from applying individual residues (i.e., the mixture produces a non-additive
18 synergistic effect) has not been systematically tested and is currently unknown. In this study,
19 we used ¹³C isotope labelled cover crop residues (i.e., buckwheat, clover, radish, and
20 sunflower) to track the fate of plant residue-derived C and C derived from the priming of
21 SOM in treatments comprising a quaternary mixture of the residues and the average effect
22 of the four individual residues one day after residue incorporation in a laboratory

23 microcosm experiment. Our results indicate that, despite all treatments receiving the same
24 amount of plant residue-derived C (1 mg C g⁻¹ soil), the total microbial biomass in the
25 treatment receiving the residue mixture was significantly greater, by 26% (3.69 μg C g⁻¹),
26 than the average microbial biomass observed in treatments receiving the four individual
27 components of the mixture one day after applying crop residues. The greater microbial
28 biomass C in the quaternary mixture, compared to average of the individual residue
29 treatments, that can be attributed directly to the plant residue applied was also significantly
30 greater, by 132% (3.61 μg C g⁻¹). However, there was no evidence that the mixture resulted
31 in any more priming of native SOM than average priming observed in the individual residue
32 treatments. The soil microbial community structure, assessed using phospholipid fatty acid
33 (PLFA) analysis, was significantly ($P < 0.001$) different in the soil receiving the residue
34 mixture, compared to the average structures of the communities in soil receiving four
35 individual residues. Differences in the biomass of fungi, general bacteria, and Gram-positive
36 bacteria were responsible for the observed synergistic effect of crop residue mixtures on
37 total microbial biomass and residue-derived microbial biomass, especially biomarkers 16:0,
38 18:2ω6 and 18:3ω3. Our study demonstrates that applying a mixture of crop residues
39 increases soil microbial biomass to a greater extent than would be expected from applying
40 individual residues and that this occurs either due to faster decomposition of the crop
41 residues or greater carbon use efficiency (CUE), rather than priming the decomposition of
42 native SOM. Therefore, growing crop polycultures (e.g., cover crop mixtures) and
43 incorporating mixtures of the resulting crop residues into the soil could be an effective
44 method to increase microbial biomass and ultimately C stocks in arable soils.

45 **Key words:** diverse, mixture, synergistic, ¹³C-PLFA, crop residues

46 **1. Introduction**

47 Soil organic carbon (SOC) plays a critical role in global carbon (C) dynamics in the earth
48 system and is a major property influencing soil functions and health. Applying crop residues
49 to soils is a common strategy used in agroecosystems to enhance SOC stocks (Chapman and
50 Newman, 2010; Chen et al., 2015). When microorganisms decompose plant litter and use
51 the C for metabolism, they catabolise a portion of this C, which is usually released as carbon
52 dioxide (CO₂), and simultaneously assimilate and anabolise a portion of C into their biomass.
53 After their death, microbial necromass returns to soil and contributes to SOC storage (Liang
54 et al., 2020). However, applying fresh crop residues could also stimulate microbial
55 decomposition of native SOM via co-metabolism or mining for nutrients, resulting in a
56 priming effect (Kuzyakov, 2010; Wang et al., 2015). A recent meta-analysis including 2048
57 individual experimental comparisons from 94 laboratory incubation studies revealed that
58 the addition of exogenous organic C significantly enhanced native SOC decomposition, by
59 47.5%, in terrestrial ecosystems, with the highest priming effect observed in arable soils
60 (60.9%) (Sun et al., 2019). Therefore, using crop residues to maximize SOC stocks requires a
61 consideration of the impact of amendments on SOC priming.

62 Previous studies have investigated the C mass balance after the application of plant residues
63 of a single plant species (Rubino et al., 2010; Shahbaz et al., 2018). However, crop residues
64 returned to soils under arable land management practices such as intercropping, rotations,
65 and cover crops include the residues of more than one plant species. Based on studies
66 examining decomposition dynamics in soils receiving mixed species residues, it cannot
67 necessarily be assumed that the application of a crop residue mixture will have the same
68 impact on C dynamics as predicted from observations made on the impact of the individual

69 residues (Gartner and Cardon, 2004; Porre et al., 2020). If the behaviour or effect of a
70 residue mixture can be predicted from the behaviour of the individual residues, this is
71 classified as an additive effect (Redin et al., 2014). By contrast, a mixture could also deliver
72 an antagonistic non-additive effect (i.e. the mixture's effect is less than the average of
73 individual species) or a synergistic non-additive effect (i.e. the mixture's effect is greater
74 than the average of individual species), which suggests there are interactions, via microbial
75 decomposers, among the constituents of the mixture (Redin et al., 2014).

76 A majority of the previous studies have explored the effects of mixtures on litter
77 decomposition and associated C cycling and nutrient release (e.g. nitrogen mineralization
78 and immobilization) by focusing on leaf litter decomposition in forest ecosystems (Castro-
79 Díez et al., 2019; Gartner and Cardon, 2004; Mao et al., 2017). Even the most recent meta-
80 analysis on litter mixtures only focused on litter mass loss due to decomposition (Porre et al.,
81 2020). To better understand the mechanisms of plant species diversification on soil C
82 dynamics, we need to fill the knowledge gap regarding mixture effects on the microbial fate
83 of C supplied by individual components of the mixture and the potential for interactions of
84 residue C with older soil organic matter (SOM) via priming effects.

85 The mechanisms responsible for non-additive effects in residue mixture decomposition
86 dynamics are not fully understood and might depend strongly on the context within which
87 the study was conducted (Porre et al., 2020). To explain non-additive effects, processes
88 relating to nutrient transfer between nutrient-rich (low C:N) and nutrient poor (high C:N)
89 litters, transfer of inhibitory compounds from one species' litter to another, or physical
90 (water retention) effects have been frequently mentioned (Porre et al., 2020). In addition,
91 mixing chemically contrasting crops may provide a greater number of niches for

92 microorganisms to exploit, which allows functionally dissimilar microbial communities to
93 coexist, and thus result in a greater microbial diversity and biomass than might be expected
94 from the average of the individual communities that are supported by monocultures
95 (Chapman and Newman, 2010). Although greater diversity not always linked to greater
96 biomass, a recent meta-analysis found that in temperate regions, where soil is not rich in C,
97 greater microbial diversity is usually associated with greater microbial biomass due to
98 facilitation and niche partitioning by supporting the co-existence of multiple microbial
99 species (Bastida et al., 2021). An increased microbial diversity could increase the probability
100 of including taxa particularly influential in extracellular enzyme production for nutrient
101 mining and SOM degradation through metabolic and co-metabolic processes. Furthermore,
102 an increased microbial biomass could also result in stronger SOM mineralization as a
103 consequence of increased microbial metabolic activity (Bastida et al., 2021). Thus, mixtures
104 of crop residues could enhance the extent to which soil microbial communities mine
105 nutrients for SOM and lead to a greater priming effect. By contrast, some studies showed
106 that mixtures of diverse plant residues have a greater opportunity to provide preferable
107 growth substrates for microbes (e.g., if the average C/N ratio of residues is close to 24),
108 thereby offsetting the extent to which SOM decomposition is primed and decreasing
109 microbial anabolism of primed SOM (Xiao et al., 2015). These two mechanisms (i.e.,
110 increasing or offsetting priming effect) could occur simultaneously and combine to
111 determine the overall magnitude of the priming effect.

112 In this study, we investigated the residues of four functionally dissimilar crops from four
113 different plant families (i.e., buckwheat, clover, radish, and sunflower) which are widely
114 grown in mixtures as cover crops in agricultural systems. We established a microcosm

115 experiment comprising treatments receiving either mixtures or individual (non-mixture) ^{13}C
116 labelled cover crop residues which provided the same amount of residue-derived C (1 mg C
117 g^{-1} soil). Soil phospholipid fatty acid (PLFA) analysis was undertaken one day after
118 incorporating crop residues to quantify the biomass of key soil microbial groups. Gas
119 chromatography-combustion-stable isotope mass spectrometry (GC-C-IRMS) was used to
120 identify the microbial groups that had incorporated residue-derived C and, by mass balance,
121 quantify the amount of primed SOM-derived C which was incorporated into the microbial
122 biomass. The difference between the mixture and the average of four non-mixtures enabled
123 us to determine whether the mixture delivered either a synergistic (mixture > average), an
124 antagonistic (mixture < average), or an additive (mixture = average) effect.

125 We assumed that microbes have no preference for ^{13}C over ^{12}C . Because a mixture of crop
126 residues may increase the niche breadth and provide a more diverse supply of nutrients,
127 thereby creating conditions enhancing the growth and facilitation of multiple co-exist
128 microorganisms. Thus, we hypothesised that the mixture could result in a synergistic effect
129 on total microbial biomass. Compared to individual species, the mixture has higher
130 probability to provide preferable growth substrates which could be more easily assimilated
131 into microbial biomass, and thus we would expect a synergistic effect of the mixture on the
132 microbial biomass derived from plant residues. Given that the cover crop species tested had
133 divergent C:N ratios (ranging between 10 and 32 and spanning the threshold C:N (≈ 24) for
134 net N mineralization- immobilization (Norton and Schimel, 2011), we hypothesised that
135 adding residues in a mixture (average C/N = 17) would decrease the requirement for
136 microorganisms to prime native SOM to scavenge for N and therefore induce an
137 antagonistic effect on the microbial biomass C derived from primed SOM.

138 **2. Materials and methods**

139 **2.1. Soil samples and crop residues**

140 A silty loam Luvisol (World Reference Base classification); pH (H₂O) 6.3, 22.32 g C kg⁻¹, 2.24 g
141 N kg⁻¹, 0.90 mg NH₄⁺ -N kg⁻¹, 2.75 mg NO₃⁻ -N kg⁻¹ was collected from an arable field on the
142 University of Reading's research farm at Sonning, Reading, UK (51.481152, -0.902188) in
143 August 2019 after harvesting spring barley (*Hordeum vulgare*). Seven surface soil samples
144 (0-20 cm depth) were randomly sampled and mixed thoroughly to create one homogenous
145 sample, approximately 20 kg in weight.

146 Four cover crops, buckwheat (*Fagopyrum esculentum*), berseem clover (*Trifolium*
147 *alexandrinum*), oil radish (*Raphanus raphanistrum*), and sunflower (*Helianthus annuus*),
148 were continuously and uniformly labelled with ¹³C₂O₂ in growth chambers by IsoLife
149 (Wageningen, Netherlands). Buckwheat and clover were harvested 5 weeks after sowing,
150 while radish and sunflower were harvested after 4 weeks. The ¹³C atom percent of the
151 resulting aboveground biomass was 6.7%, 7.8%, 7.8%, and 8.0% for buckwheat, clover,
152 radish, and sunflower, respectively. Corresponding unlabelled crops were grown under the
153 same conditions in growth chambers by IsoLife (Wageningen, Netherlands), and harvested
154 at the same time. After harvesting, the aboveground residues of both ¹³C labelled and
155 unlabelled crops were dried at 70 °C and milled to pass through 0.05 mm mesh. The
156 chemical composition of ¹³C labelled and unlabelled residues is provided in Table S1.

157 **2.2. Experimental design**

158 Soil was sieved to pass a 4 mm mesh and then pre-incubated for 7 days at 26 °C with a soil
159 water content of 60% of the water holding capacity (0.22 g g⁻¹). As indicated in Table 1, the
160 treatments consisted of pure unlabelled residues, labelled non-mixture residues, quaternary

161 mixtures of residues which contained one labelled species and three unlabelled species, and
162 a control without any crop residue additions. For each treatment receiving residues, four
163 replicate microcosms were established by mixing 150 g of fresh soil (equivalent to 122.95 g
164 dry soil) thoroughly with a mass of dry residues to ensure C was added to each microcosm
165 at a rate of 1 mg C g⁻¹ soil. In the pure treatments, all the added C was from the same
166 unlabelled residue sample. In the non-mixture treatments, 25% of added C was from the ¹³C
167 labelled residue and 75% of added C was from the unlabelled residue of the same crop
168 species. In the mixture treatments, 25% of added C was from a ¹³C labelled crop and 75% of
169 added C comprised unlabelled residues from the other three crop species.

170 For the measurement of soil respiration, a 100 g subsample was transported from each
171 microcosm to a bulk density ring (98 cm³), stored in a gas-tight plastic jar (365 cm³), and
172 kept open during incubation. The rest of sample was kept in an opened plastic bag for the
173 measurement of PLFA. Both measurements of PLFA and soil respiration were taken one day
174 after applying cover crop residues.

175 To measure soil respiration, jars were sealed with a Suba-Seal® Septa for 1 h and a 16 ml
176 headspace gas sample was taken from each jar using a syringe and hypodermic needle,
177 transferred into pre- evacuated vials, and analysed with gas chromatography (Agilent 7890B,
178 UK) (Adekanmbi et al., 2020). The universal gas law was used to determine the amount of
179 CO₂ (ng g⁻¹ soil h⁻¹) emitted from each jar.

180 **Table 1 Experimental design and the C/N ratio of added crop residues in each treatment**

Treatment	Plant	Abbreviation	Description	Added ¹³ C amount (mg C g ⁻¹ soil)	C/N ratio
Non-mix	Buckwheat	NB	25% labelled + 75% unlabelled buckwheat	0.0168	10
Non-mix	Clover	NC	25% labelled + 75% unlabelled clover	0.0195	32
Non-mix	Radish	NR	25% labelled + 75% unlabelled radish	0.0195	18
Non-mix	Sunflower	NS	25% labelled + 75% unlabelled sunflower	0.0200	19
Mix	Buckwheat	MB	25% labelled buckwheat, 75% of unlabelled residues (clover, radish, sunflower)	0.0168	17
Mix	Clover	MC	25% labelled clover, 75% unlabelled residues (buckwheat, radish, sunflower)	0.0195	17
Mix	Radish	MR	25% labelled radish, 75% unlabelled residues (buckwheat, clover, sunflower)	0.0195	17
Mix	Sunflower	MS	25% labelled sunflower, 75% unlabelled residues (buckwheat, clover, radish)	0.0200	17
Pure	Buckwheat	PB	%100 unlabelled buckwheat	\	9
Pure	Clover	PC	%100 unlabelled clover	\	30
Pure	Radish	PR	%100 unlabelled radish	\	21
Pure	Sunflower	PS	%100 unlabelled sunflower	\	22
Soil			Soil only without any plant residue addition	\	\

181 *Note: 25% and 75% refers to proportion of total added C. The quantity of C added was the same across all treatments apart from the soil treatment, which*
 182 *was 1 mg C g⁻¹ soil. The C/N ratio in the mixture treatments were the average of four types of mixture treatments.*

183

184 **2.3. Phospholipid-derived fatty acids (PLFA) extraction**

185 One day after incubation with soil and crop residues, a 10 g aliquot of soil was sampled from
186 each replicate plastic bag and freeze-dried for downstream analysis of PLFA. PLFA was
187 extracted following the method described by Sizmur et al. (2011). Briefly, 4 g of freeze-dried
188 soil was extracted with 7.8 ml of Bligh and Dyer extractant containing chloroform: methanol:
189 citrate buffer (1:2:0.8 v/v/v). The extracted phospholipids were methanolized as fatty-acid
190 methyl esters and dissolved in hexane for analysis by gas chromatography (GC).

191 **2.4. Gas chromatography (GC)**

192 PLFA methyl esters were analysed using an Agilent Technologies 6890N gas chromatography
193 equipped with a Supercowax 10 capillary GC column (60 m × 0.25 mm i.d. × 0.25 µm film
194 thickness) and a Flame Ionisation Detector (FID). Helium was the carrier gas. The
195 temperature programme was 1-minute isothermal at 60 °C, followed by a ramp to 145 °C at
196 25 °C per minute, followed by an increase to 250 °C at 2.5 °C per minute and then held
197 isothermally at 310 °C for 10 minutes. Data were processed using GC ChemStation (Agilent
198 Technologies). Peaks were identified using a bacterial fatty acid methyl esters (BAME) mix
199 (Sigma Aldrich, UK) and quantified using a 37-component fatty acid methyl esters (FAME)
200 mix (Sigma Aldrich, UK). The biomass of each group of microorganisms was determined
201 using the combined mass of fatty acids to which the group is attributed in Table S2.

202 **2.5. Gas chromatography- combustion-isotope ratio mass spectrometry (GC-C-IRMS)**

203 GC-C-IRMS analysis were performed by injecting a 1 µl sample of fatty-acid methyl esters
204 into an Agilent 7890N GC, upstream of a DELTA V™ Isotope Ratio Mass Spectrometer
205 (electron ionization, 100 eV, 1 mA electron energy, 3 F cup collectors m/z 44, 45 and 46,
206 CuO/Pt Thermofisher GC IsoLink interface maintained at 1000°C). A Nafion membrane was

207 employed to prevent water from reaching the ion source. GC conditions were the same as
208 that described above. Samples were calibrated against reference CO₂ of known isotopic
209 composition, which was introduced directly into the source five times at the beginning and
210 end of every run. Data was processed in the Isodat Gas Isotope Ratio MS Software
211 (ThermoFisher Scientific) to generate $\delta^{13}\text{C}$ values representing the ratio of $^{13}\text{C}/^{12}\text{C}$ in fatty-
212 acid methyl esters, relative to the $^{13}\text{C}/^{12}\text{C}$ ratio of the international Pee Dee Belemnite (PDB)
213 standard (0.01118).

214 $\delta^{13}\text{C}$ values obtained for methylated compounds were corrected for the addition of
215 derivative C using the mass balance equation following equation 1 to 3 (Zhang et al., 2019):

$$\delta^{13}C_{PLFA}(\text{‰}) = [(N_{PLFA} + 1) \times \delta^{13}C_{FAME} - \delta^{13}C_{MeOH}] / N_{PLFA} \quad (\text{Eq 1})$$

$$R_{\text{sample}} = (\delta^{13}C_{PLFA} / 1000 + 1) \times R_{PDB} \quad (\text{Eq 2})$$

$$F = R / (R + 1) \quad (\text{Eq 3})$$

216

217 where $\delta^{13}C_{PLFA}$, $\delta^{13}C_{FAME}$, and $\delta^{13}C_{MeOH}$, were the $\delta^{13}\text{C}$ value of PLFA, FAME and methanol
218 ($\delta^{13}C_{MeOH}$ was assumed to equal zero), respectively. N_{PLFA} was the number of C atoms in this
219 fatty acid to which the fatty acid methyl ester corresponds. R_{sample} and R_{PDB} were the ratios
220 of $^{13}\text{C}/^{12}\text{C}$ in sample and in PDB. F is the fractional isotopic abundance representing the
221 concentration of ^{13}C as a proportion of the total concentration of C in the PLFA.

222 Equation 4 was used to calculate how much ^{13}C ($\mu\text{g } ^{13}\text{C g}^{-1}$ soil) was incorporated into PLFA
223 biomass in the non-mixture treatments:

13C incorporation ($\mu\text{g } ^{13}\text{C} / \text{g soil}$) in non – mixture (Eq 4)

$$= F_{\text{non-mixture}} \times C_{\text{non-mixture}} - F_{\text{pure}} \times C_{\text{pure}}$$

224 $F_{\text{non-mixture}}$ and F_{pure} were the fractional abundances of ^{13}C in a non-mixture treatment its
225 corresponding pure unlabelled treatment, respectively. $C_{\text{non-mixture}}$ and C_{pure} were the
226 concentration of PLFAs ($\mu\text{g C g}^{-1}$ soil) in the non-mixture and its corresponding pure
227 unlabelled treatment, which were the results of GC-FID analysis. For example, ^{13}C
228 incorporation in the non-mixture of buckwheat (NB) equals to $F_{\text{NB}} \times C_{\text{NB}} - F_{\text{PB}} \times C_{\text{PB}}$

229 For each mixture treatment, there was 75% of added C from three different unlabelled
230 residues (25% from each residue). The resulting ^{13}C incorporation ($\mu\text{g } ^{13}\text{C g}^{-1}$ soil) of these
231 unlabelled residues was accounted for using equation 5.

13C incorporation ($\mu\text{g } ^{13}\text{C} / \text{g soil}$) in mix (Eq

$$= F_{\text{mix}} \times C_{\text{mix}} - 0.25 \times (F_{\text{PB}} \times C_{\text{PB}} + F_{\text{PC}} \times C_{\text{PC}} + F_{\text{PR}} \times C_{\text{PR}} + F_{\text{PS}} \times C_{\text{PS}}) \quad 5)$$

232 where F_{mix} was the fractional abundances of ^{13}C in mixture treatment. F_{PB} , F_{PC} , F_{PR} , F_{PS} , were
233 the fractional abundances of ^{13}C in pure unlabelled buckwheat, clover, radish, and sunflower
234 treatment, respectively. C_{mix} , C_{PB} , C_{PC} , C_{PR} , C_{PS} , were the concentration of PLFAs ($\mu\text{g C g}^{-1}$ soil)
235 in the treatments of mixture, pure buckwheat, pure clover, pure radish, and pure sunflower,
236 which were analysed by GC-FID.

237 The proportion (unitless) of ^{13}C incorporated from total amount of added ^{13}C was calculated
238 using equations 6 and 7:

$$\begin{aligned} \text{Total amount of added } ^{13}\text{C} (\mu\text{g } ^{13}\text{C} / \text{g soil}) & \quad (\text{Eq} \\ & = \text{Total added C} \times \text{Proportion of } ^{13}\text{C} \text{ labelled residue} \quad 6) \\ & \times ^{13}\text{C atom percent of labelled residue} \end{aligned}$$

$$\text{PLFA proportion of added } ^{13}\text{C} = \frac{\text{ } ^{13}\text{C incorporation}}{\text{Total amount of added } ^{13}\text{C}} \quad (\text{Eq} \\ 7)$$

239 where total added C was 1000 $\mu\text{g C g}^{-1}$ soil, the proportion of ^{13}C labelled residue was 0.25.
240 The ^{13}C atom percent was 6.7%, 7.8%, 7.8% and 8.0% for buckwheat, clover, radish, and
241 sunflower residues, respectively. This brings the total amount of added ^{13}C to 16.75, 19.50,
242 19.50, and 20.00 $\mu\text{g } ^{13}\text{C g}^{-1}$ soil, for treatments which received 25% of their added C from ^{13}C
243 labelled buckwheat, radish, clover, and sunflower.

244 ^{13}C is a tracer to track the amount of C derived from crop residues and incorporated into
245 PLFA biomass. Assuming microorganisms have no preference for ^{12}C over ^{13}C , the proportion
246 of PLFA incorporated ^{13}C from added ^{13}C should also be the same to the proportion of PLFA
247 incorporated ^{12}C from total added ^{12}C . Following this logic, in all the non-mixture treatments,
248 the amount of C incorporated into PLFA derived from added crop residue C was calculated
249 using equation 8:

$$\begin{aligned} \text{PLFA derived from residues in non - mixture } (\mu\text{g C/g soil}) & \quad (\text{Eq} \\ & = \text{PLFA proportion of added } ^{13}\text{C} * 1000 \mu\text{g C/g soil} \quad 8) \end{aligned}$$

250 where 1000 $\mu\text{g C g}^{-1}$ soil was the total added amount of C (equal to 1 mg C g^{-1} soil).

251 In the mixture treatment, the ^{13}C labelled residue contributed 25% of the added C (i.e., 250
252 $\mu\text{g C g}^{-1}$ soil). The proportion of ^{13}C in each type of mixture (calculated by equation 7) was

253 multiplied by 250 ($\mu\text{g C g}^{-1}$ soil) to get the amount of PLFA derived from this ^{13}C labelled crop.
254 We then added up the amount of PLFA derived from each type of ^{13}C labelled crop to form
255 the PLFA which was derived from the mixture of crop residues (equation 9).

$$\begin{aligned} & \text{PLFA derived from residues in mixture } (\mu\text{g C/g soil}) && \text{(Eq 9)} \\ & = (\text{Proportion}_{MB} + \text{Proportion}_{MC} + \text{Proportion}_{MR} + \text{Proportion}_{MS}) && 9) \\ & * 250 \mu\text{g C/g soil} \end{aligned}$$

256 where proportion_{MB} , proportion_{MC} , proportion_{MR} , and proportion_{MS} represent the
257 proportion (unitless) of ^{13}C incorporated from added ^{13}C in four types of mixture treatments
258 where each type was ^{13}C -labelled by buckwheat, clover, radish, and sunflower, separately.

259 The PLFA derived from primed SOM was calculated as the difference between the total PLFA
260 (measured using GC-FID), PLFA derived from residues, and PLFA in the control soil without
261 any residue addition using equation 10.

$$\begin{aligned} & \text{PLFA derived from SOM priming } (\mu\text{g C/g soil}) && \text{(Eq 10)} \\ & = \text{total PLFA} - \text{PLFA derived from residues} - \text{PLFA in the soil only treatment} \end{aligned}$$

262 where total PLFA and PLFA in the soil only treatment were analysed by GC-FID and
263 converted to $\mu\text{g C g}^{-1}$ soil.

264 For mixture treatment, because there were only four samples have crop residue-derived
265 PLFA, to calculate SOM-derived PLFA, we created a general mixture of which total PLFA was
266 the average of total PLFA in four different types of mixture in the same replicate as
267 calculated by equation 11.

$$\begin{aligned} & \text{Total PLFA in mixture } (\mu\text{g C/g soil}) && \text{(Eq 11)} \\ & = 0.25 * (\text{total PLFA}_{MB} + \text{total PLFA}_{MC} + \text{total PLFA}_{MR} + \text{total PLFA}_{MS}) \end{aligned}$$

268 Therefore, there were four replicates for total PLFA, residue derived PLFA, and SOM derived
269 PLFA in the general mixture.

270 **2.6. Data analysis**

271 All the statistics were conducted in R (version 3.5.2) (R Core Team, 2018) except for the
272 analysis of similarity (ANOSIM) which was conducted in Primer 7 (Primer-e, Auckland, New
273 Zealand). We used nested analysis of variance (ANOVA) to compare the effect from the
274 quaternary mixtures with the effect from the average of the four non-mixture treatments
275 on total PLFA biomass, the PLFA biomass incorporated from crop residues, and the PLFA
276 biomass incorporated from primed SOM.

277 To compare the difference in microbial community structure between the mixture and the
278 average of four individuals, PLFA data measured by GC-FID was “Hellinger” transformed in
279 all treatments. Non-metric multidimensional scaling (NMDS) on Bray-Curtis distance of the
280 transformed data was performed using the “vegan” package in R (Oksanen et al., 2019).
281 Bray-Curtis distance similarity matrices were analysed by a one-way analysis of similarity
282 (ANOSIM) using Primer 7 to test if the differences between the mixture and the average of
283 four non-mixture treatments were significant.

284 **3. Results**

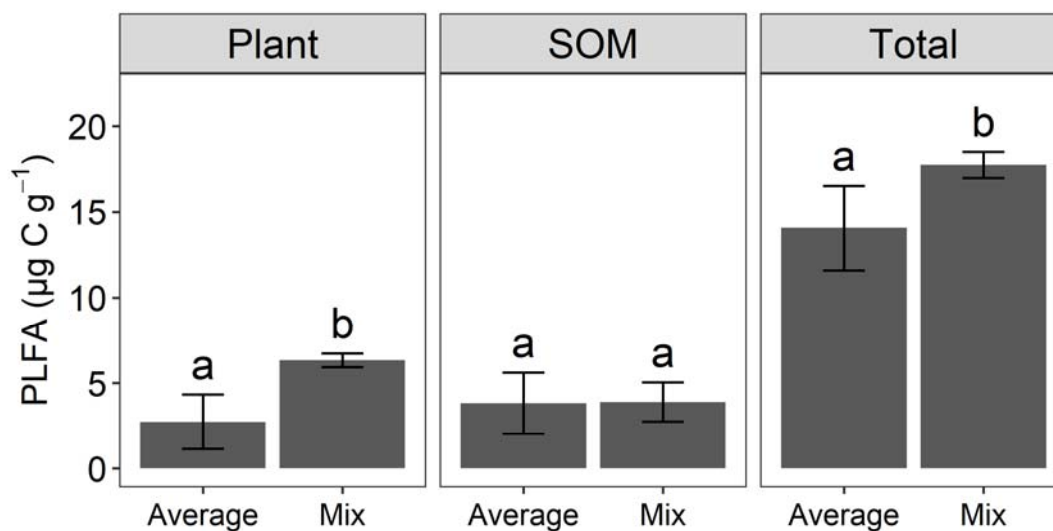
285 **3.1 Crop residues increased soil microbial biomass and altered community structure**

286 The incorporation of cover crop residues significantly ($P < 0.001$) shifted the microbial
287 community structure away from the unamended control soil (Figure S1). The incorporation
288 of residues from different crop species lead to significantly ($P < 0.05$) different community
289 structures (Figure S1). The biomass of general bacteria, Gram-positive bacteria, and fungi in
290 the unamended control soil was 1.80, 2.20, and 2.36 $\mu\text{g C g}^{-1}$, respectively, which was

291 greater than the biomass of Gram-negative bacteria and protozoa (Table S3). Despite the
292 same rate of C addition applied across all the treatments, total PLFA biomass differed
293 between residue amendment treatments; ranging from 11.66 to 18.93 $\mu\text{g C g}^{-1}$, which was
294 significantly ($P < 0.05$) greater than that in the control soil (Table S3).

295 3.2 Mixing crop residues increased total PLFA biomass and altered microbial 296 community structure

297 Total PLFA biomass was 17.74 and 14.05 $\mu\text{g C g}^{-1}$ for the mixture and the average of four
298 non-mixture treatments, respectively (Figure 1). The soils in the mixture treatments had a
299 significantly ($P < 0.05$) greater total microbial biomass, by 26% (3.69 $\mu\text{g C g}^{-1}$), compared to
300 the soils in the non-mixture treatments (Figure 1 and Table 2).



301

302 **Figure 1** Total PLFA biomass (Total), PLFA biomass incorporated from crop residues (Plant),
303 and the PLFA biomass incorporated from soil organic matter (SOM) in the mixture and the average
304 of four non-mixture treatments. Error bars are standard deviations. Different letters within the
305 same panel indicate a significant difference between treatment means at $P < 0.05$.

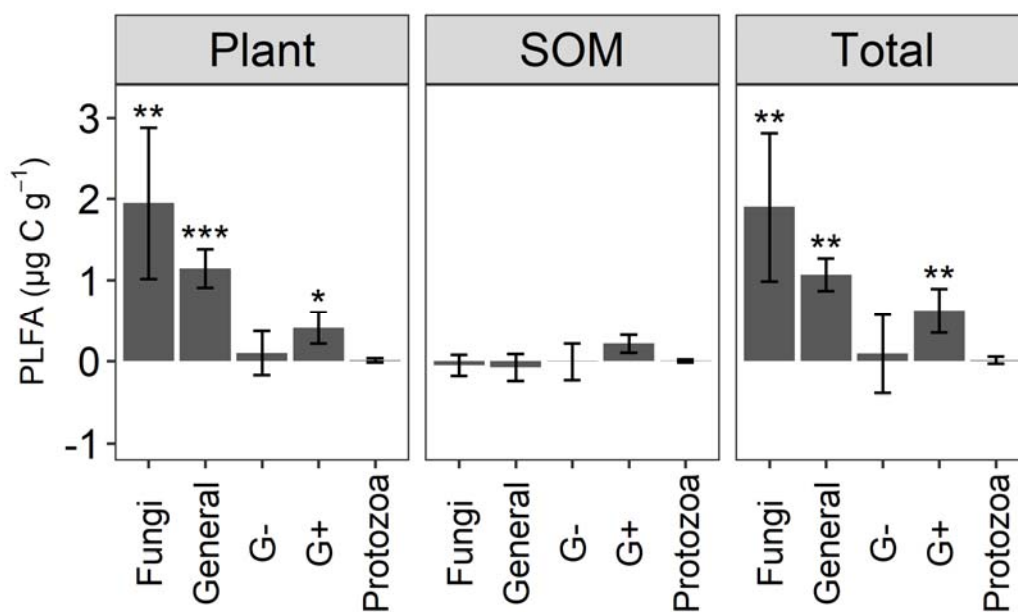
306

307 **Table 2** NEST ANOVA (treatment nested with crop species) to compare the effects of mixture and the species of plant on total PLFA, crop residue-
 308 derived PLFA, and SOM (soil organic matter)- derived PLFA at different microbial groups. MIX has two levels (mix or non-mix), crop species have five
 309 levels (radish, clover, buckwheat, sunflower, and mixture). DF is the degree of freedom. Values are F value. *, **, and *** represent significant level at P
 310 < 0.05 , 0.01 , and 0.001 . G+ and G- bacteria are Gram-positive and Gram-negative bacteria, respectively.

Total PLFA									
	DF	Sum of PLFA	General bacteria	G+ bacteria	G- bacteria	Fungi	Protozoa	G+/G-	Fungi/Bacteria
Treatment	1	15.16**	9.16**	12.54**	0.07	9.66**	0.12	0.17	1.80
Treatment:Species	3	5.78**	2.13	9.55***	4.32*	2.69	0.95	3.46*	4.01*
Crop residue- derived PLFA									
Treatment	1	20.62***	16.93***	4.64*	0.19	11.43**	0.22	0.40	0.02
Treatment:Species	3	1.36	0.12	1.34	2.66	1.41	0.85	0.71	0.4
SOM- derived PLFA									
Treatment	1	0.01	0.07	1.33	0.00	0.03	0.01	0.44	0.53
Treatment:Species	3	2.72	2.14	3.97*	3.53*	1.48	0.81	1.08	2.63

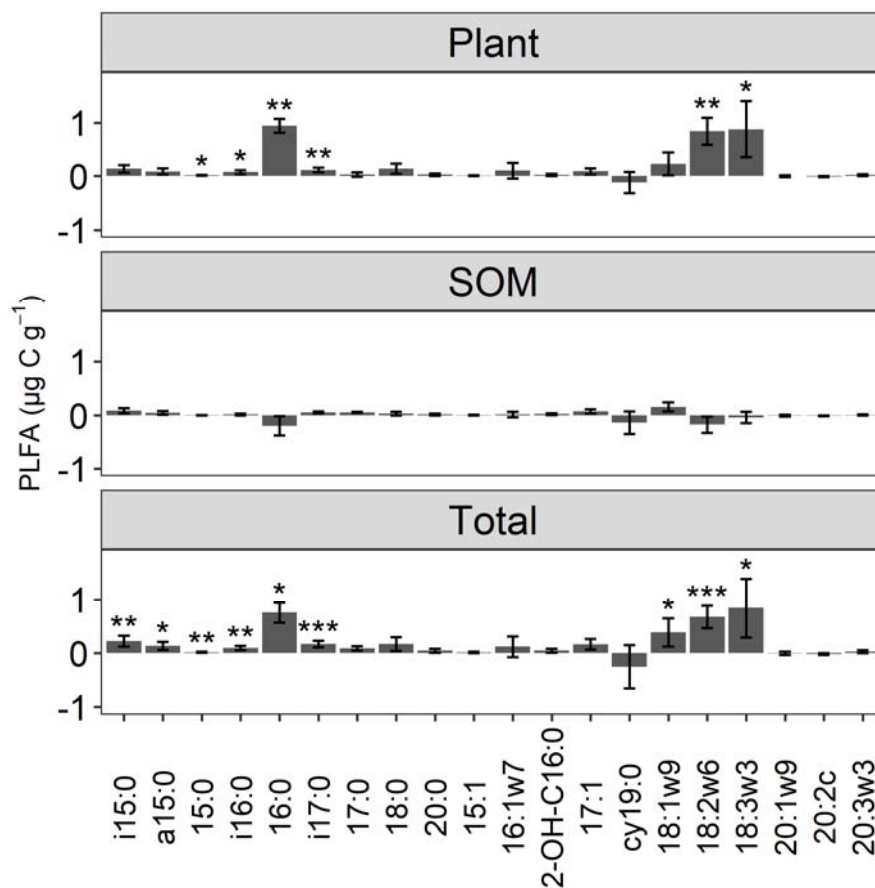
311

312 The mixture treatment resulted in a significantly ($P < 0.001$) greater general bacteria, Gram-
313 positive bacteria, and fungi biomass by 31% ($1.07 \mu\text{g C g}^{-1}$), 18% ($0.62 \mu\text{g C g}^{-1}$), and 38%
314 ($1.90 \mu\text{g C g}^{-1}$), respectively, than the average of the non-mixture treatments (Figure 2). In
315 particular, biomarkers i15:0, 16:0, 18:1 ω 9, 18:2 ω 6 and 18:3 ω 3 were significantly ($P < 0.05$)
316 more abundant in the mixture treatment than the average of four non-mixture treatments
317 (Figure 3). The ratios of fungi to bacteria in the mixture and the average of the four non-
318 mixture treatments were 0.64 and 0.57, which was not significantly different (Figure S2).
319 Similarly, there was no significant difference in the ratio of Gram-positive to Gram-negative
320 bacteria between the mixture and the average of four non-mixture treatments (Figure S3).



321

322 **Figure 2** Differences in PLFA biomass of key microbial groups between the mixture and the
323 average of four non-mixture treatments. Positive value means mixture treatment has greater
324 biomass than the average of non-mixture treatments. The three panels represent the crop derived
325 PLFA biomass (Plant), the primed SOM derived PLFA biomass (SOM), and the total PLFA biomass
326 (Total). Error bars are standard deviations. General, G+, and G- represent general bacteria, Gram-
327 positive and Gram-negative bacteria, respectively. *, **, or *** means significantly different from
328 zero at the level $P < 0.05$, 0.01 or 0.001 .

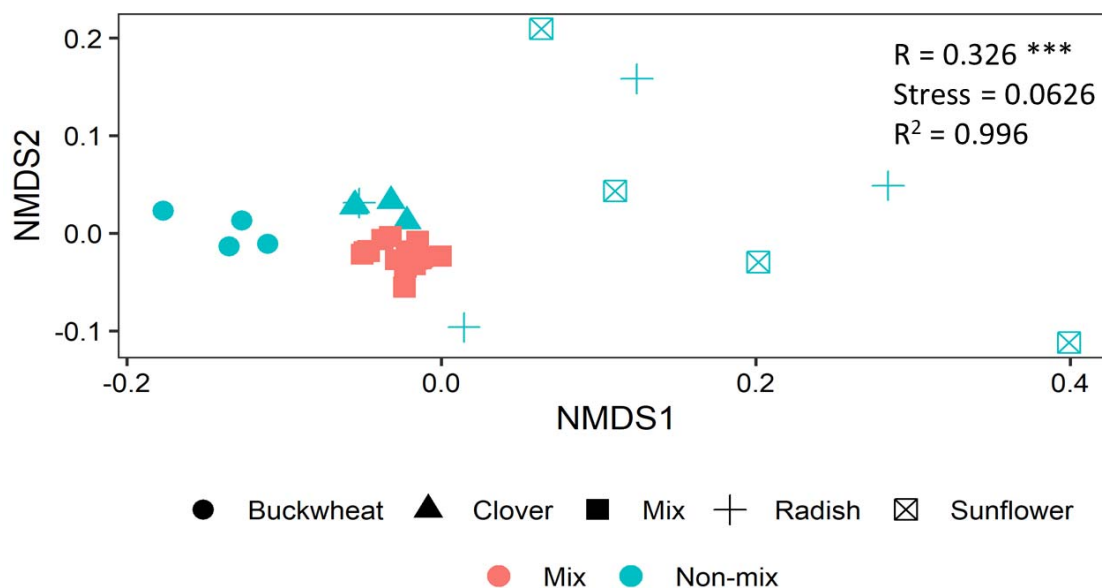


329

330 **Figure 3** Differences in PLFA biomarker biomass between the mixture and the average of
331 four non-mixture treatments. Positive values indicate that mixture has a greater biomass than the
332 average of four non-mixtures. The three panels represent the crop residue derived PLFA biomass
333 (Plant), the primed SOM derived PLFA biomass (SOM), and the total PLFA biomass (Total). Error
334 bars are standard deviations. *, **, and *** indicates significantly different from zero at the level P
335 < 0.05 , 0.01 , and 0.001 , respectively.

336

337 One-way ANOSIM results demonstrated that microbial community structure in the mixture
338 was significantly ($P < 0.001$, $R = 0.326$) different from that in the average of four non-
339 mixture (Figure 4).



340

341 **Figure 4** *Microbial community structure in the mixture and non-mixture treatments. Non-*
342 *metric multidimensional scaling (NMDS) on the Bray-Curtis distance on the Hellinger transformed*
343 *PLFA data. Each symbol represents one sample. In the non-mixture treatment, different shapes*
344 *represent different crop species. R value followed by “***” indicates significant ($P < 0.001$)*
345 *difference between the mixture and the average of four non-mixture treatments analysed by one-*
346 *way ANOSIM.*

347

348 3.3 Mixing crop residues increased the PLFA biomass derived from crop residues

349 Applying ^{13}C labelled cover residues allowed us to distinguish the microbial biomass derived
350 from plant residues from other resources. The results showed that application of crop
351 residues in the mixture and the average of four non-mixture treatments resulted in 6.34 and
352 $2.73 \mu\text{g C g}^{-1}$ PLFA biomass derived from crop residues, respectively (Figure 1). We observed
353 a significantly ($P < 0.05$) greater crop residue derived PLFA, by 132% ($3.61 \mu\text{g C g}^{-1}$), in the
354 mixture treatment, compared to the average of four non-mixture treatments (Figure 1).

355 The mixture exhibited a significantly ($P < 0.01$) greater biomass of general bacteria, Gram-
356 positive bacteria, and fungi derived from crop residues, by 193% ($1.15 \mu\text{g C g}^{-1}$), 86% (0.41
357 $\mu\text{g C g}^{-1}$), and 158% ($1.95 \mu\text{g C g}^{-1}$), respectively, compared to the average of four non-

358 mixture treatment (Figure 2 and Table 2). The biomass of 16:0, 18:2 ω 6, and 18:3 ω 3 in the
359 mixture treatment were significantly ($P < 0.05$) greater than those in the average of non-
360 mixture treatments (Figure 3). Based on the differences in PLFA biomass derived from crop
361 residues, there was no significant differences in either the fungi to bacteria ratio, or the
362 Gram-positive to Gram-negative bacteria ratio, between mixture and the average of four
363 non-mixture treatments (Figure S2 and S3).

364 **3.4 Mixing crop residues did not increase the PLFA biomass derived from primed SOM**

365 The PLFA biomass derived from primed SOM was the difference between total PLFA, PLFA in
366 the control soil, and the PLFA derived from plant residues which was tracked by ^{13}C . The
367 results showed that the PLFA biomass derived from primed SOM was 3.90 and 3.82 $\mu\text{g C g}^{-1}$
368 for the mixture and the average of four non-mixture treatments, respectively (Figure 1).
369 Therefore, we observed only 2% (0.08 $\mu\text{g C g}^{-1}$) greater PLFA biomass derived from primed
370 SOM between the mixture and the average of four non-mixture treatments which was not
371 statistically significant (Figure 1 and Table 2). None of the microbial groups or the
372 corresponding PLFA biomarkers exhibited significant differences between the mixture and
373 the average of four non-mixtures in terms of SOM priming (Figure 2 and 3). There were no
374 significant differences in either the fungi to bacteria ratio or the Gram-positive to Gram-
375 negative bacteria of PLFA derived from primed SOM between the mixture and the average
376 of four non-mixture treatments (Figure S2 and S3).

377 **4. Discussion**

378 Following the addition of ^{13}C -labelled residues, we compared the effect of mixing crop
379 residues with the effect of applying the residues of a single crop species, on the soil
380 microbial community composition and attributed the source of the C assimilated by the

381 microbial biomass. Applying crop residues as a mixture resulted in a significantly ($P < 0.05$)
382 greater microbial biomass compared to the average effect of applying each of the residues
383 individually, indicating a synergistic effect of crop residue diversity on soil microbial biomass
384 (Figure 1). This result was consistent with a microcosm experiment which found 38.2%
385 higher soil microbial biomass N in soils receiving a mixture of residues than would be
386 expected from residues of individual species (Mao et al., 2017). The mixture also exhibited
387 greater total microbial biomass than any of the non-mixture treatments (Figure S4),
388 suggesting its synergistic effect is more likely to be caused by facilitation between residues
389 rather than by the disproportionate effect of individual species. The common explanation in
390 the literature for why a mixture induces synergistic effect on decomposition rate is that N is
391 transferred from low C/N residues to high C/N residues to satisfy microbial stoichiometric
392 requirement (Gartner and Cardon, 2004; Mao et al., 2017). If the availability of N is the
393 limiting factor in our experiment, we would see the treatment receiving clover residues
394 (which had the highest C/N ratio) exhibiting the smallest microbial biomass. On the contrary,
395 the total microbial biomass in the treatment receiving clover residue (where C/N ratio was
396 32) was significantly ($P < 0.05$) greater than the treatment receiving buckwheat residue
397 (where C/N ratio was 10) (Figure S4), implying that N transfer between high C/N and low
398 C/N residues to satisfy microbial stoichiometric requirement may not be the reason for the
399 observed synergistic effects of the mixture.

400 Although crop residues, such as buckwheat and sunflower in our study, contain high amount
401 of N, they could also contain a considerable amount of plant secondary metabolites (e.g.,
402 tannins and terpenes) that suppress microbial resource assimilation (Gessner et al., 2010) or
403 require specialized enzymes to be degraded (Chomel et al., 2016). When mixing residues,

404 microorganisms could acquire energy via decomposing labile C from other residues to
405 synthesize enzymes to degrade these secondary metabolites, and then anabolise
406 subsequent monomers in their biomass (Wang et al., 2015). Additionally, micronutrients
407 (e.g., Mg and Ca) were identified to play a paramount role in regulating nutrient transfer
408 between residues (García-Palacios et al., 2016). This includes positive effects of Ca on the
409 growth and activity of white rot fungi and significant effects of Mg on invertebrate (e.g.,
410 nematodes) diets which may influence the soil microbial community composition (García-
411 Palacios et al., 2016). Thus, we suggested that the non-additive effect induced by the
412 mixture was not predominantly controlled by residue bulk elemental composition (C/N
413 ratio); instead, it could have been driven by the chemical composition of residues, including
414 plant secondary metabolites and micronutrients.

415 We found that the greater microbial biomass in soils receiving residue mixtures, compared
416 to individual residues, can largely be attributed to C assimilated directly from the plant
417 residues, rather than C obtained by enhanced priming of SOM (Figure 1). Microbial
418 communities preferentially mineralise labile compounds after crop residues are applied to
419 soils to build their biomass rather than decomposing pre-existing SOM (Ball et al., 2014).
420 Significantly greater crop residue-derived C was observed in biomarkers 16:0, 18:2 ω 6 and
421 18:3 ω 3 in the soils amended with crop residue mixtures, compared to soils receiving
422 individual residues (Figure 3). This finding is supported by the knowledge that biomarker
423 16:0 represents microbial groups that primarily build their biomass from crop derived C
424 (Wang et al., 2014). Although 16:0 was widely accepted as general bacterial biomarker, it
425 also found in crops (Willers et al., 2015). There is a possibility that this biomarker was
426 present as part of the tissues of our cover crops. Therefore, future studies combining stable

427 isotope probing and DNA fingerprints would overcome this uncertainty associated with the
428 PLFA approach (Blagodatskaya and Kuzyakov, 2013). We found that fungi (e.g., biomarker
429 18:2 ω 6 and 18:3 ω 3) were particularly efficient in assimilating a mixture of crop residue-
430 derived C (Figure 3). This could be because fungal hyphae networks allow nutrients to be
431 transported between microsites in the chemically and spatially heterogenous environment
432 resulting from mixed residues (Ball et al., 2014). Furthermore, fungi can produce a wide
433 range of extracellular enzymes that can degrade compounds which are recalcitrant for other
434 microbes (Voriskova and Baldrian, 2013). By contrast, Gram-negative bacteria (e.g., 16:1 ω 5
435 and 16:1 ω 7) are good at incorporating SOM derived C into their biomass (Nottingham et al.,
436 2009).

437 Although the application of crop residues did induce microbial assimilation of the native
438 SOM, the magnitude of assimilation was not significantly changed by mixing crop residues
439 (Figure 1 and 2). This observation was contrary to a previous assertion that mixtures of plant
440 residues with a wide spectrum of labile compounds may support a higher microbial biomass
441 and produce more extracellular enzymes, consequently enhancing the potential to prime
442 the decomposition of recalcitrant compounds in SOM (Meier and Bowman, 2008). The lack
443 of a marked mixture-induced priming effect could, however, be because the amount of C
444 added was the same in all treatments. A recent meta-analysis, which analysed studies
445 applying a range of organic C application rates, up to 3 mg C g⁻¹, reported that the
446 magnitude of the priming effect significantly increased with the increasing rate of additions,
447 but was not affected by different residue types (Sun et al., 2019).

448 We made our observations only one day after crop residues were applied to soils because
449 we were interested in identifying the microbial groups which incorporate plant derived C

450 directly into their biomass by anabolism, rather than secondary turnover of C from microbial
451 necromass, which could continue for months or years (Gunina and Kuzyakov, 2015). If
452 incubating longer until the exhaustion of available C, then mixture treatments which have a
453 larger microbial biomass, may facilitate a stronger priming effect because of the increased
454 need for nutrients to maintain microbial survival (Yu et al., 2020).

455 Our study revealed that the incorporation of crop residue mixtures increased the total
456 microbial biomass during the first day after application to a greater extent than would be
457 expected by the addition of the same quantity of C from a single species residue. This
458 additional microbial biomass is mostly derived from the crop residues themselves, rather
459 than primed SOM. It is not clear whether the reason for this greater microbial biomass is
460 due to faster metabolism of the residues, or higher carbon use efficiency (CUE) of the
461 microorganisms responsible. We also found that soil respiration rates one day after applying
462 crop residues were not significantly ($P < 0.05$) different between the mixture and the
463 average of four non-mixtures (Figure S7). If the mechanism of higher CUE is responsible (or
464 even partially responsible), then applying a mixture of crop residues to soils may help to
465 increase SOC stocks in arable soils because more crop residue-derived C enters the microbial
466 biomass and less is respired as CO₂ (Figure S7) and the larger microbial biomass may result
467 in a larger microbial necromass and, ultimately, more SOM. Therefore, land management
468 practices that involve the incorporation of a mixture of crop residues (such as growing cover
469 crop mixtures in polyculture after monoculture crop harvest and before planting the
470 following monoculture crop) may be beneficial for increasing soil C stocks and climate
471 change mitigation.

472 **5. Conclusion**

473 Mixing of crop residues produced a synergistic effect on total soil microbial biomass because
474 fungi, general bacteria, and Gram-positive bacteria were able to incorporate plant-derived C
475 directly into their biomass to a greater extent than when applying individual crop residues.
476 These findings may be due to a better balance of nutrient inputs provided by mixtures, and
477 broader niche for microorganism to colonise. Residue addition primed native SOM, but
478 mixing residues resulted in an additive effect on microbial biomass C resulting from primed
479 SOM. If the greater soil microbial biomass from residue mixtures results in greater microbial
480 necromass and does not result in later priming of native SOM, then applying a mixture of
481 crop residues to soils (e.g., incorporating or growing a cover crop polyculture) could be a
482 sustainable agricultural practice to enhance SOC storage in arable soils.

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491 **Supplementary information**

492 A Supplementary information document is provided which contains 3 tables and 7 figures

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