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Understanding the combined impacts of weeds and climate change on crops

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3 32 Inés Ibáñez 0000-0002-1054-0727
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5 33 **Classification**

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7 34 Biological Sciences, Ecology.
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9 35 **Keywords**

10 36 Drought, elevated CO₂, meta-analysis, non-native plants, plant competition, warming
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12 37 **Author Contributions**

13
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15 39 A.T. performed research; I.I. and MV analyzed data; M.V. and I.I. wrote the paper; E.B., D.B.,
16 40 B.A.B., R.E., B.B.L., A.T., J.D. and C.J.B.S. commented several versions of the manuscript.
17 41

18 42 **This PDF file includes:**

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49 Abstract

50 Crops worldwide are simultaneously affected by weeds, which reduce yield, and by climate
51 change, which can negatively or positively affect both crop and weed species. While the
52 individual effects of environmental change and of weeds on crop yield have been assessed, the
53 combined effects have not been broadly characterized. To explore the simultaneous impacts of
54 weeds with changes in climate-related environmental conditions on future food production, we
55 conducted a meta-analysis of 171 observations measuring the individual and combined effects of
56 weeds and elevated CO₂, drought or warming on 23 crop species. The combined effect of weeds
57 and environmental change tended to be additive. On average, weeds reduced crop yield by 28 %,
58 a value that was not significantly different from the simultaneous effect of weeds and
59 environmental change (27%), due to increased variability when acting together. The negative
60 effect of weeds on crop yield was mitigated by elevated CO₂ and warming, but added to the
61 negative effect of drought. The impact of weeds with environmental change was also dependent
62 on the photosynthetic pathway of the weed/crop pair and on crop identity. Native and non-native
63 weeds had similarly negative effects on yield, with or without environmental change. Weed impact
64 with environmental change was also independent of whether the crop was infested with a single
65 or multiple weed species. Since weed impacts remain negative under environmental change, our
66 results highlight the need to evaluate the efficacy of different weed management practices under
67 climate change. Understanding that the effects of environmental change and weeds are, on
68 average, additive brings us closer to developing useful forecasts of future crop performance.

69

70 Main text

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72 Introduction

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74 As the human population grows, global demand for food production is increasing. Concurrently,
75 factors affecting food supply are changing. The spread of weed species and the prevalence of
76 herbicide-resistant weeds is increasing. Weeds already cause greater global crop losses than
77 either insect pests or pathogens (Oerke 2006, Fried *et al* 2017); yield losses to non-native weeds
78 can amount to 42% of crop production (Vilà *et al* 2004). Weed control costs farmers over €150
79 million per year in the UK (Williamson 2002) and \$3 billion per year in the U.S. (Pimentel *et al*
80 2005). Simultaneously, changes in Earth's climate and atmosphere are directly affecting growing
81 conditions for plants; colder regions are experiencing longer growing seasons (Mueller *et al*
82 2015), drought conditions are increasing in many regions (Naumann *et al* 2018), and rising
83 atmospheric CO₂ is affecting plant growth worldwide (Zhu *et al* 2016). Some of these changes
84 are causing widespread yield losses in crops (Porter *et al* 2014). For example, in South Asian
85 smallholder farms, drought and other water constraints cause yield losses that average 9.1% in
86 wheat, rice, sorghum and chickpea crops (Li *et al* 2011). Furthermore, a recent meta-analysis of

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3 87 studies modeling climate warming impacts on crops found that models project yield losses of
4 88 wheat, rice and maize to increase in tropical and temperate regions in the second half of the
5 89 century (Challinor *et al* 2014). However, most predictions of future crop yields are based solely on
6 90 crop performance under forecasted climates without accounting for changes in weed competition.

9 91 Combined effects of climate change and weeds on crop production have not been broadly
10 92 synthesized, but have important implications for future crop management practices (Thomson *et*
11 93 *al* 2010). A primary question is whether the combined effect of weeds and climate change is
12 94 additive (individual effects sum together), synergistic (effects amplify each other) or antagonistic
13 95 (effects offset each other) (Crain *et al* 2008, Darling and Côté 2008, Jackson 2015). Some
14 96 studies that have tested multiple abiotic global change factors have found additive effects
15 97 (Dieleman *et al* 2012). However, many of these effects are not additive (16) and interactions
16 98 between abiotic and biotic global factors can be complex (Tylianakis *et al* 2008). If non-additive
17 99 effects of climate and weeds are common, predictions of future crop yields will have to include
18 100 them to be realistic (Tubiello *et al* 2007, Ramesh *et al* 2017).

24 101 In agricultural systems, both crops and weeds are influenced by multiple climate-related
25 102 environmental conditions (Korres *et al* 2016). Changes in atmospheric CO₂, temperature and
26 103 precipitation influence weed and crop species' metabolic rates, phenology and performance
27 104 (Bunce and Ziska 2000). However, weeds and crops may respond to these changes differently
28 105 because they have been subjected to distinct selective pressures (Korres *et al* 2016). Further,
29 106 research on biological invasions suggests that the interaction between environmental change and
30 107 weed effects could depend on the functional traits of the species involved, the origin of the
31 108 weeds, and whether one or more weeds are present. For example, the impact of weeds on crops
32 109 often depends on the plants' functional traits, such as their photosynthetic pathways (Ziska 2003,
33 110 Fried *et al* 2017). Everything else being equal, increased atmospheric CO₂ increases primary
34 111 production and water-use efficiency in C3 plants, while C4 plants are less likely to benefit from
35 112 CO₂ enhancement. In contrast, C4 plants are more likely than C3 plants to thrive under warm and
36 113 dry conditions (Ainsworth and Long 2005, Prior *et al* 2011). Thus, the competitive outcome
37 114 between C3 and C4 plants could depend on the specific environmental component of climate
38 115 change under consideration (Korres *et al* 2016). Since both crops and weeds include C3 and C4
39 116 plants, we expect that impacts on crop yield will depend on interactions between photosynthetic
40 117 pathway and environmental change (Ainsworth and Rogers 2007).

49 118 Effects of weeds on crops might also depend on weed origin (native versus non-native). Non-
50 119 native plants have left behind natural enemies that keep their populations in check in their native
51 120 ranges (Maron and Vilà 2001). Release from natural enemies can allow non-native plants to
52 121 allocate more resources to growth and reproduction in the new regions, and become more
53 122 competitive (Blossey and Notzold 1995). Many successful non-native plant species also have

123 broad environmental tolerances, high phenotypic plasticity or the ability to evolve more rapidly
124 than native plants (Davidson *et al* 2011, Simberloff *et al* 2012) potentially allowing them to benefit
125 more from global environmental change than native plants (Davidson *et al* 2011). Thus, with
126 environmental change, we expect non-native weeds to have greater impacts on crop yield than
127 native weeds.

128 The magnitude of weed impacts on crops under environmental change might also depend on
129 whether a crop is infested by one or multiple weed species. Most crops contain diverse
130 communities of weeds, which respond to environmental change through shifts in relative
131 abundance (Booth and Swanton 2002). Ecological theory and empirical evidence suggest that a
132 community of multiple species could be more resilient to environmental change than poor species
133 communities (Tilman *et al* 2014, van der Plas 2019). Thus, in agricultural systems, we expect
134 infestation by multiple weed species to have greater impacts on crop yield under environmental
135 change.

136 Understanding the interactive effects of climate and weeds requires empirical studies that
137 compare crop yields under different environmental conditions in the presence of weeds
138 (Parmesan *et al* 2018). Some experiments have tested these effects, but these studies have yet
139 to be synthesized quantitatively. As a result, we do not have clear expectations for how climate
140 change and weeds will affect crops, simultaneously. To test the above hypotheses and identify
141 the contexts in which crop yield is most vulnerable to the simultaneous effects of weeds and
142 environmental change, we conducted a systematic review and meta-analysis. Specifically, we
143 analyzed results from experiments addressing the combined and direct effects of weeds and
144 elevated CO₂, drought or warming on the yield-related variables of 23 crop species, and asked
145 the following questions: (i) Is the effect of weeds on crop yield altered by environmental change?
146 (ii) Are the combined effects of weeds and environmental change on crops additive, synergistic or
147 antagonistic? (iii) Do the combined effects of weeds and environmental change depend on the
148 photosynthetic pathway (C3 vs. C4) of the crop/weed species pair, (iv) on the origin of the weed
149 (native vs. non-native), or (v) on whether single or multiple weed species are competing with the
150 crop? Finally, (vi) how might the main crop species around the world be affected by weeds under
151 environmental change?

152 **Materials and Methods**

153 ***Literature search and data selection criteria***

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156 Our database development was based on a systematic literature search protocol, paper selection
157 criteria and data extraction protocol (Pullin and Stewart 2006). For quality control, at each step,
158 we trained data collectors using an example subset of the data and discussed eligibility of all
159 included data.

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3 160 To identify studies that experimentally tested the interactive effects of weeds and climate changes
4 161 (elevated CO₂, warming or drought), we searched the Web of Science core collection for all
5 162 records until 25/07/2018 using the following keywords: (i) “crop AND (weed control OR herbicide
6 163 OR weed competition OR weed management) AND weed AND (Warm* OR heat* Or thermal OR
7 164 temperature increase OR temperature manipulation* OR climate change)”; (ii): “crop AND (weed
8 165 control OR herbicide OR weed competition OR weed management) AND weed AND (CO₂ OR
9 166 carbon dioxide) AND (increase* OR enhance* OR enrich* OR elev*)”; and (iii) “crop AND (weed
10 167 control OR herbicide OR weed competition OR weed management) AND weed AND (Drought
11 168 OR water stress* OR rainout OR rain out OR rain-out OR precipitation exclusion* OR rain
12 169 exclusion* OR precipitation removal*)”.

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18 170 This search retrieved 1,436 publications. By reviewing titles and abstracts, we identified studies
19 171 for which the following criteria for data inclusion were met: (i) the study independently tested the
20 172 effects on crop performance of both the weed and environmental change ; (ii) the study tested the
21 173 combined effects of the weed and environmental change either through experimental
22 174 manipulation of both factors, or by experimentally manipulating one factor across a gradient of the
23 175 other factor (e.g. a weed removal experiment across an irrigation gradient); (iii) the study included
24 176 control treatments (no weed and no environmental change); and finally; (iv) the response
25 177 variables were measured simultaneously in all treatments. These criteria for inclusion yielded a
26 178 set of 57 publications (SI References, Figure S1).

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31 179 A single publication could include results of multiple observations. If the publication reported
32 180 results fitting our criteria for data inclusion for multiple weed and/or crop species, we considered
33 181 each weed-crop combination to be a unique observation. If several varieties of the same crop
34 182 were tested independently, we also considered these to be unique observations. If an article
35 183 included observations conducted on the same crop but located in two or more regions or sites,
36 184 we considered the studies as independent. Similarly, if the treatments were conducted several
37 185 times, or the crop was planted at different times, each treatment was used as an independent
38 186 observation. When the observation incorporated information on more than one control treatment
39 187 (e.g. different herbicides used to suppress weeds), we included them as independent
40 188 observations. Following the same reasoning, when the article incorporated information on more
41 189 than one experimental method for the same environmental change variable (e.g. CO₂ enrichment
42 190 conducted in both growth chambers and a field experiment), we considered each separately.
43 191 When more than two treatment levels were examined (e.g. different weed densities, different CO₂
44 192 concentrations), only the most extreme treatment was included. Thus, if the degree of weed
45 193 infestation varied, we compared the effects of the lowest (“control”) vs. highest (“treatment”) level
46 194 of infestation.

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3 195 Studies reported different crop response variables (e.g. plant biomass, seed production, plant
4 196 height, leaf area, etc.). We considered the response variable most associated to the specific crop
5 197 yield (crop yield hereafter). If the response variables were measured several times, we provided
6 198 the average value of the time series. If the time series was not provided, we included the
7 199 measure that we considered ecologically most representative (e.g. the last one in the time series;
8 200 spring measurement of an annual series during season of maximum activity; measurement
9 201 closest to maximum crop yield).

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14 202 For every unique observation, we recorded the weed species and the location of the observation,
15 203 using this information to determine whether the weed was native or non-native to the study region
16 204 based on range information provided in several information sources (e.g. CABI Invasive Species
17 205 Compendium). We also recorded whether the observation focused on a single weed or a mixture
18 206 of weeds. Crop and weed species were also classified by their photosynthetic pathway (C3 vs.
19 207 C4).

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23 208 The analysis included field, greenhouse and chamber experiments. Weed treatments were rather
24 209 heterogeneous. Weed treatments used included: planting weeds at different densities, removing
25 210 weeds manually or mechanically, use of herbicides or combinations of these removal methods. In
26 211 field conditions, drought has mostly tested by different irrigation treatments or by comparing wet
27 212 and dry seasons or years. Similarly, the effect of warming was tested in experiments that
28 213 elevated soil or air temperature but also in studies that compared years with different mean
29 214 temperature but similar precipitation. The effect of increased CO₂ included similar numbers of
30 215 studies in outdoor open-top chambers as in indoor chambers.

31 32 33 34 35 216 **Data analysis**

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37 217 We examined the effect of weeds and environmental change using standard meta-analytical
38 218 models (Koricheva *et al* 2013). For each observation, we extracted data on the number of
39 219 replicates, mean and variability around the mean (e.g. standard deviation or standard error) for
40 220 controls, individual treatments, and interactive weed and environmental change treatments. We
41 221 used the Web Plot Digitizer online application (<http://arohatgi.info/WebPlotDigitizer/app/>) to
42 222 extract values from figures in the papers. When empirical data were not presented, or were
43 223 presented only in summarized format, we emailed corresponding authors to request raw data and
44 224 included any raw data received in the analysis. A description of the flowchart for the publication
45 225 selection process following Moher *et al* (2009) can be found in Figure S1.

46 47 48 49 50 51 226 *Effect size calculation*

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53 227 We compared treatment effects across cases by estimating effect size (ES) as: $In (Treatment$
54 228 $mean/Control mean)$. We used simulations (1,000 iterations) to estimate ES mean and SD, ES for
55 229 each observation was drawn from normal distributions with reported means and SDs (see

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3 230 Supplemental Information for code). Effect size was estimated at each iteration and from that
4 231 output (1,000 values) we estimated ES mean and SD (SI text S1). Sample size was also
5 232 considered in these estimations by weighing reported variances by sample size (Gurevitch and
6 233 Hedges 2001). We used simulations to estimate ES, instead of standard metrics (e.g. Hedges' g)
7 234 because a large proportion of observations did not report a measure of variability associated with
8 235 the mean (57%). We included these observations by estimating the variance around their ES as a
9 236 latent variable.

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14 237 Although there is a lack of consistency about how to handle missing variance data (Wiebe *et al*
15 238 2006), there are three common methods of dealing with this: an algebraic calculation which
16 239 requires parametric summary statistics, trial-level imputation (averaging, or running regressions,
17 240 across observations with known variances), and no imputation (excluding observations with no
18 241 variance) (Batson and Burton 2016). We did not want to bias our results by excluding such a
19 242 large proportion of the data, as a result, in our analyses, we opted for the most conservative,
20 243 lowest bias, imputation method. We estimated the missing variances as a function of the largest
21 244 ES variance calculated from observations with reported variances. We sampled from normal
22 245 distributions (limited to be positive) with estimated largest variance as the mean and a SD of 1.
23 246 There were also nine observations that did not report sample size. For these observations, we
24 247 followed the most conservative approach and assigned them a sample size N=1.

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30 248 We calculated the expected additive effect of weeds plus environmental change by summing the
31 249 individual experimental results (Weed+ Environmental Change) and compared the expected
32 250 additive effect to the measured combined effect reported in each observation (Weed&
33 251 Environmental Change). We followed Jackson (2015) to estimate the mean and pooled SD of the
34 252 additive effect (see ES4 in Table 1).

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38 253 To address our specific research questions, we calculated several effect sizes (ES), all based on
39 254 crop performance under different treatments, C: control, W: with weeds, EC: under environmental
40 255 change, and W&EC: with weeds and under environmental change (Table 1, SI text S1).

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264 **Table 1. Calculations of effect size (ES) estimates** to assess the combined effects of
 265 environmental change and weeds on crop performance. C: control, crop performance without
 266 weeds & without environmental change; W: crop performance under weed treatment & without
 267 environmental change; EC: crop performance under environmental change & without weeds;
 268 W&EC: observed crop performance with weeds and environmental change; W+EC: expected
 269 additive performance (i.e. sum of the individual experimental results) with weeds and
 270 environmental change.

Effect size	Comparison	Calculation
ES1	Weed effect on crops under current climatic conditions	$\ln(W/C)$
ES2	Environmental Change (elevated CO ₂ , drought or warming) effect on crops	$\ln(EC/C)$
ES3	Observed weed effect under Environmental Change	$\ln(W\&EC/EC)$
ES4	Additive expectation relative to the observed combined effect	$\ln(W+EC/W\&EC)$

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272 *Analysis of effect sizes*

273 Individual values of effect size were then analyzed to assess effects of weeds and environmental
 274 change factors on crop production. Effect sizes were analyzed using mixed effects models with
 275 publication as a random effect. This accounted for the lack of independence among observations
 276 from the same study. By using study random effects in our analyses, individual observations were
 277 nested within each study, thus the study random effect is a 'combined' mean, as in Ponisio *et al*
 278 (2015). Given the low number of studies considering the combined effects of weeds and
 279 environmental change on crops, including other potential random effects (e.g., for crop and weed
 280 species) was not feasible. For each ES calculation the effect of different environmental change
 281 factors (elevated CO₂, drought or warming) were estimated. Since we were using latent estimates
 282 of effect size variability (for those observations with missing variance), we used a hierarchical
 283 Bayesian approach in this analysis; parameters were all estimated from non-informative prior
 284 distributions except for the missing variances (see description of methods above). All the prior
 285 distributions for the effect size were: ES*~Normal (0,100), and all the precision terms prior
 286 distributions were: 1/variance~Gamma (0.001, 0.001).

287 We ran similar analyses, using ES1-effect of weeds alone and ES3-weed effect under
 288 environmental change, for each combination of crop/weed photosynthetic pathway (C3 and C4),
 289 for each crop species, for each type of weed origin (native and non-native), and for single vs.
 290 multiple weed species systems. Due to the low number of data points for some of the subgroups,
 291 this second analysis was done without publication random effects. Effect size calculations and
 292 analyses were carried out in OpenBUGS (Thomas *et al* 2006; see SI text S1 for analysis code).

293 Effect size posterior estimates that did not include zero in their 95% credible intervals were
 294 considered statistically significant. Effect sizes with 95% credible intervals that did not overlap
 295 were considered significantly different from each other.

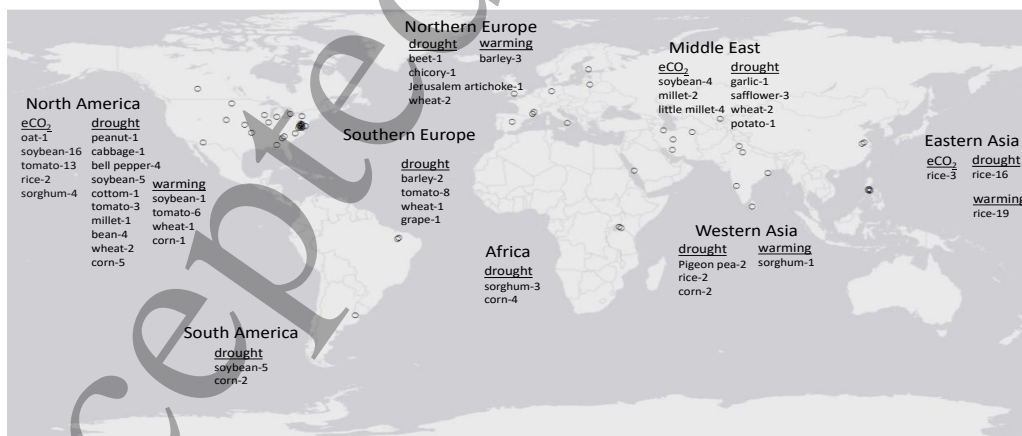
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297 *Publication bias*

298 Meta-analysis results may be distorted by publication bias, that is, the selective publication of
 299 articles finding significant effects over those that find non-significant effects (Rothstein 2008). In
 300 our case, this bias in publication could lead to an overestimate of the effects of weeds and
 301 environmental change variables on crop yield. We visually checked for potential bias using funnel
 302 plots (see Figure S2; although see Tang and Liu 2000, Lau *et al* 2006).

303 **Results and Discussion**

304 Our final database contained 171 observations from 57 publications (Table S1) on the effect of
 305 more than 47 weed species on 23 crop species. Most observations were conducted in North
 306 America (72) and Asia (44), followed by Europe (Fig. 1), with a clear lack of observations
 307 conducted in Africa, South America, and Australasia. The majority of observations (84) were on
 308 the effect of drought, with 49 on the effect of elevated CO₂ and 31 on the effect of increased
 309 temperature. The most frequently studied crops were rice (42 observations), mostly in Eastern
 310 Asia, followed by soybean (31), tomato (30) and corn (12). Wheat, the most widely grown crop in
 311 the world and second most important food source in low-income countries, was represented in
 312 only seven observations, none of them testing the effects of elevated CO₂. Nine crop species
 313 were represented by a single observation (Fig. 1).



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315 **Figure 1.** Geographic distribution of study sites used in the analysis. Tables show crop species
 316 studied and environmental factor considered. Numbers indicate number of observations included
 317 in the meta-analyses (See Table S1 for more information).

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3 319 *Is the effect of weeds on crop yield altered by environmental change?*
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5 320 Weeds alone significantly reduced crop performance by 27.99 % on average (Fig. 2, Table S3).
6 321 Elevated CO₂ increased crop performance by 45.90% while drought decreased it by 29.85%;
7 322 warming did not have a significant effect due to its large variation across studies (Fig. 2, Table
8 323 S3). Elevated levels of atmospheric CO₂ often increase growth and water use efficiency of crop
9 324 species that translate to increased crop production (Ainsworth and Rogers 2007). On the
10 325 contrary, drought can have devastating effects to crop yields especially in non-irrigated systems
11 326 (Li et al. 2011). The effect of warming is more context-dependent. Warming can accelerate and
12 327 improve growing conditions in temperate regions by lengthening growing seasons and periods of
13 328 time with optimal temperature but can also increase the risk of exposure to damaging heat
14 329 (Tubiello *et al* 2007).

15 330 We assessed whether the negative effects of weeds were likely to change with environmental
16 331 change by comparing ES1 (weed effect under current environmental conditions) and ES3 (weed
17 332 effect under environmental change). Overall, the simultaneous effect of weeds and environmental
18 333 change reduced crop performance by 26.64% a value that was not significantly different from the
19 334 single effect of weeds without environmental change. Crop yield became more variable with
20 335 warming, such that there was no significant effect under increased temperature (Fig. 2). ES3 was
21 336 dependent on the biome (SI text S1, Table S2). The effect of the weeds under drought was most
22 337 negative in Mediterranean, arid or semiarid climates, intermediate in temperate climates and the
23 338 lowest in tropical and subtropical climates. This indicates that the impact of weeds on major crops
24 339 might be exacerbated in dry regions such as the Mediterranean biome where models predict
25 340 decreasing precipitation with climate change (Rojas *et al* 2019). In contrast, the effect of the
26 341 weeds under warming was negative in tropical climates but not significant in temperate climates.

27 342 These results shed some light on how the simultaneous effects of environmental change on crop
28 343 and weed species may alter their interaction. Weed species tend to have a strong, positive
29 344 response to elevated atmospheric CO₂ (Ziska 2003), and weed presence counteracted any
30 345 benefits of elevated CO₂ to crops. In the case of drought, the lack of change in overall weed
31 346 impact suggests that reduced water availability has a similar negative effect on both crops and
32 347 weeds, despite the fact that the impact is larger in water stressed regions. In the case of
33 348 warming, ES3 was highly variable. A correlation analysis between the magnitude of the change
34 349 and ES3 indicated an increase in the effect size with increasing temperature differences (SI Fig.
35 350 S3). Both crops and weeds are likely to benefit from warming, leading to both positive and
36 351 negative outcomes on the impact of weeds. Overall, our results suggest more variable effects of
37 352 weeds on crops under environmental change, and a need to adapt weed management practices
38 353 where weed impacts increase (Peters *et al* 2014).
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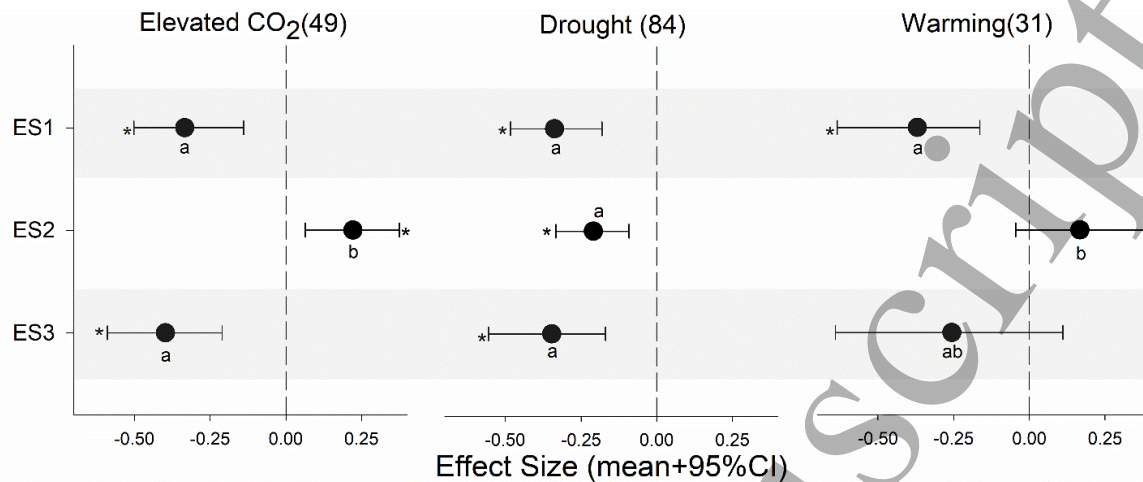


Figure 2. Effect size (ES) estimates comparing crop performance under current environment conditions and without weeds (control) with weeds (ES1), with environmental change (ES2), and the effect of weeds under environmental change (ES3). Credible intervals (95%CI) that do not include zero are considered statistically significant (indicated by an asterisk). Within each environmental change factor, different letters indicate that credible intervals are statistically different from each other. Numbers indicate sample sizes. See Table S2 for parameter values.

380 *Is the combined effect of weeds and environmental change on crops additive, synergistic or*
 381 *antagonistic?*

382 To answer this question, we compared the additive expectation against the observed combined
 383 effect of weeds and environmental change (ES4). There is wide variation among observations
 384 (Fig. 3). A correlation analysis between the magnitude of the change and ES4 indicated a trend
 385 towards synergistic effects with increasing temperature differences (SI Fig. S4). However, the
 386 combined effects of environmental change and weeds are on average additive. The effects of
 387 weeds are similar in present and predicted future environmental conditions, even though
 388 environmental change can dramatically alter competitive interactions among weeds and crops
 389 within particular cropping systems (Tylianakis *et al* 2008, Ziska and Dukes 2011). This result is in
 390 line with the additive effects found between other global change drivers (Wu *et al* 2011), but see
 391 Dieleman *et al* 2012). To realistically assess future crop production and inform management, we
 392 need to consider these combined effects of environmental change and weeds. As it stands, most
 393 experimental and synthetic work aimed at predicting crop yield only accounts for one of these two
 394 factors. Understanding that the effects of environmental change and weeds are, on average,
 395 additive brings us closer to developing useful forecasts of future crop performance.

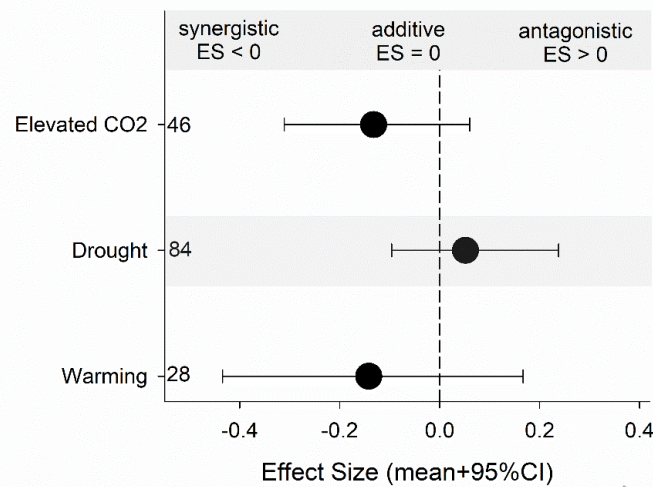


Figure 3. Effect size (ES_{4:W+EC/W&EC}) estimates to test if the combined effect of environmental change factors (EC) and weeds (W) on crop yield are additive, synergistic or antagonistic. We consider effects with credible intervals (95%) overlapping zero to be additive. Numbers are sample sizes. See Table S2 for parameter values.

411 *Does the combined effect of weeds and environmental change depend on the photosynthetic*
 412 *pathway (C3 vs. C4) of the crop/weed species pair?*

413 We addressed this question by comparing estimates between ES1 and ES3 for the four potential
 414 combinations of crop/weed photosynthetic pathway, C3/C3, C3/C4, C4/C3, and C4/C4 (SI Table
 415 S2). We found that the impact of weeds on crops grown under environmental change conditions
 416 depended on the species' respective photosynthetic pathways and on the environmental change
 417 component under consideration (Fig. 4).

418 Elevated CO₂ increased the effect of the weeds on crops if they had the same photosynthetic
 419 pathway, decreased the effect of C4 weeds on C3 crops, and was not significant in C4 crop/C3
 420 weed pairs. Thus, under elevated CO₂, weeds might increase their performance and be more
 421 competitive than crops if they are of the same photosynthetic pathway. In contrast, a greater
 422 responsiveness of C3 crops to CO₂ would benefit them when competing with C4 weeds (e.g.
 423 Patterson 1995) such as in rice crops (C3) invaded by C4 weeds (Rodenburg *et al* 2011).

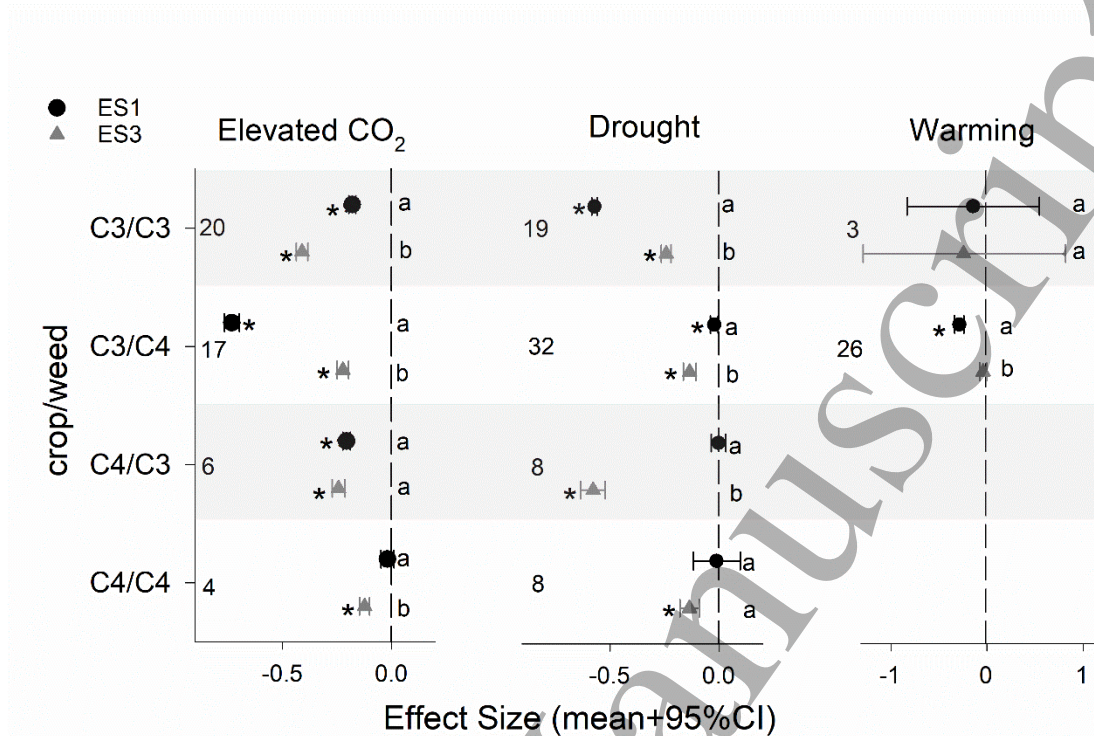


Figure 4. Effect size (ES) estimates of the effect of weeds, both under current environmental conditions (ES1) and with environmental change (ES3), categorized by weed and crop photosynthetic pathways (C3 or C4). Credible intervals (95%) that do not include zero are considered statistically significant (indicated by asterisks). For each photosynthetic pathway and environmental change factor combination, different letters indicate that the effect of the weeds is significantly different between current and changing environmental conditions (credible intervals do not overlap). There are no data to test for the effect of weeds under warming for C4/C3 and C4/C4 pairs. Numbers indicate sample sizes. See Table S2 for parameter values.

Drought increased the impact of the weeds in mixed pairs, decreased it in the C3/C3 pairs and was not significant in the C4/C4 pairs. Surprisingly, warming decreased the impact of C4 weeds on C3 crops but did not significantly affect C3 weeds' impacts on C3 crops; this combination had a small sample size (only 3 observations) and large variation. Warmer or drier conditions have been hypothesized to benefit C4 over C3 species (Patterson 1995). However, this pattern was not supported by our meta-analysis, indicating that other functional traits beside photosynthetic pathway might be more important to determine competitive superiority under climate change.

Does the combined effect of weeds and environmental change depend on the weed origin?

We addressed this question by comparing estimates between ES1 and ES3 for native and non-native weeds (SI Table S2). We did not find differences between the impact of native and non-native weeds on crops under current climatic conditions (Fig. 5). This supports other research findings that non-native plants are not more competitive than common native plants (Zhang and

van Kleunen 2019). Contrary to our expectations, environmental change did not increase the impact of non-native weeds relative to native weeds. Indeed, due to large variation across observations, non-native weeds did not consistently reduce crop performance with drought or warming. Rather, the non-native weed effects remained non-significant with environmental change (Fig. 5). This result does not align with differences found between native and non-native plant performance (i.e. survival, growth and fecundity) with climate change in natural ecosystems (Sorte *et al* 2013, Liu *et al* 2017).

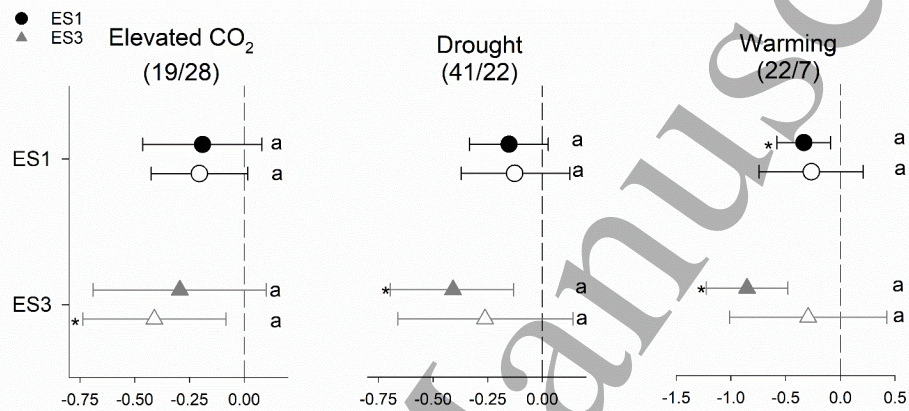
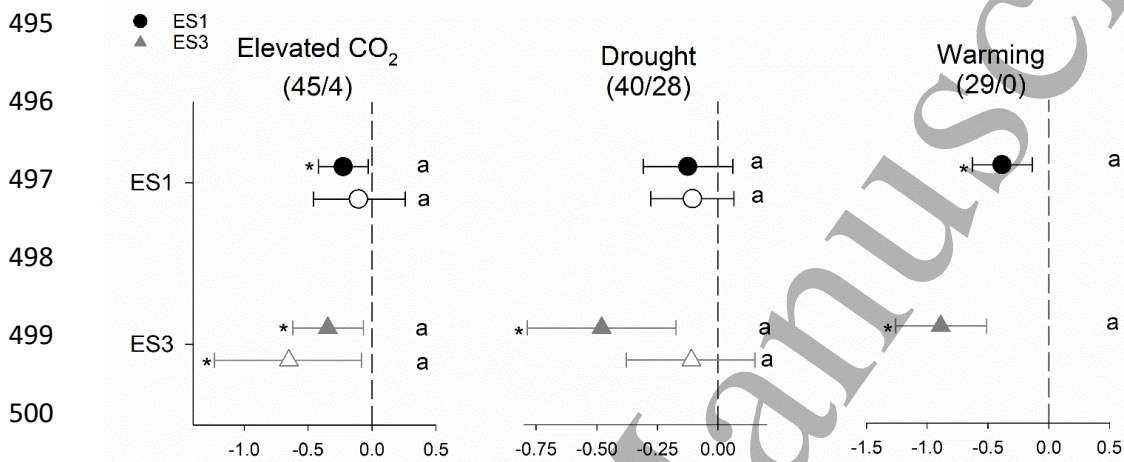


Figure 5. Effect size (ES) estimates of weeds under current environmental conditions (ES1) versus weeds with environmental change (ES3) relative to weed origin (native-solid symbols or non-native-white symbols). Credible intervals (95%) that do not include zero are considered statistically significant (indicated by asterisks). Within each panel, different letters indicate that the effects are statistically different from each other (credible intervals do not overlap). Numbers indicate sample sizes (native/non-native). See Table S2 for parameter values.

Is the combined effect of weeds and environmental change on crops similar when there is a single weed species versus multiple weed species?

We addressed this question by comparing estimates between ES1 and ES3 for single weed species versus mixtures of weeds (SI Table S2). We expected that multiple weed species would have stronger impacts on crops, and the impacts would be less affected by environmental change than single weed species. However, the impact of weeds did not differ depending on the number of weeds present, and the impact of multiple weeds was not modified by environmental change (Fig. 6). Our results suggest that the potential for diffuse competition among plant species in the community reduces the impacts on a particular species within the community (Goldberg 1987). It is possible that competition among weed species limits their impact on the crop (Lohrer and Whitlatch 2002). We also note that variability in the impacts of weed mixtures was much greater than for single weeds, particularly in the environmental change treatments.

489 While our results do not support the hypothesis that multiple weed species would have stronger
 490 impacts than a single weed, and that multiple weed species become more problematic for crops
 491 under environmental change, our sample sizes were too low to confidently reject these
 492 hypotheses, particularly under the environmental change treatments. Some studies have indeed
 493 found the reverse, that more diverse weed communities are less competitive with the crop than
 494 poor weed communities (Storkey and Neve 2018).



502 **Figure 6.** Effect size (ES) estimates of weeds under current environmental (ES1)
 503 versus weeds with environmental change (ES3) for studies with single weed
 504 species (solid symbols) vs. multiple weeds (white symbols). Credible intervals
 505 (95%) that do not include zero are considered statistically significant (indicated by
 506 an asterisk). Within each panel, different letters indicate that the effects are
 507 statistically different from each other (credible intervals do not overlap). Numbers
 508 indicate sample sizes (single/multiple). See Table S2 for parameter values.

509 *Is the effect of weeds under environmental change similar among major crop species?*

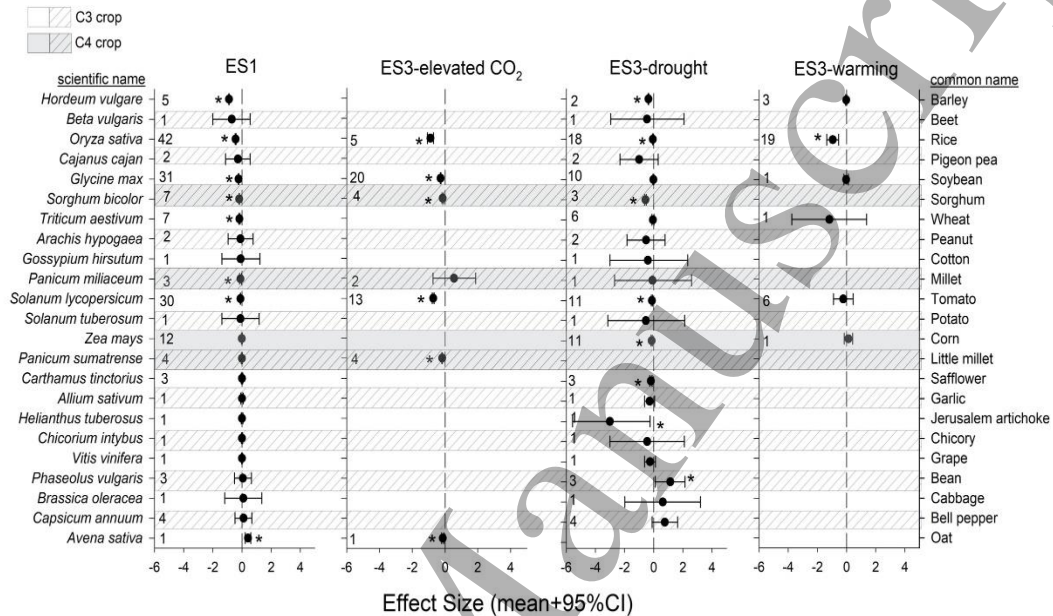
510 Ultimately, in order to effectively inform crop selection and management, we need predictions of
 511 individual crop species performance under the combined effects of weeds and environmental
 512 change. Despite the general effects of weeds and environmental change on crops (Fig. 2),
 513 individual crop species showed differing responses to environmental change (Fig. 7).

514 Surprisingly, without environmental change, crop yield was significantly reduced in only seven
 515 species, with the performance of only one crop, oat, showing an increase in yield with weeds (Fig.
 516 7, ES1). Data were available to assess the effect of weeds under elevated CO₂ for seven crops.
 517 Elevated CO₂ reversed the negative effect of weeds in millet, a C4 plant, and increased the
 518 negative effects of weeds in little millet (C4) and oat (C3).

519 The impact of weeds under drought conditions become more detrimental for corn (C4), Jerusalem
 520 artichoke (C3) and safflower (C3), and were less detrimental (non-significant) in soybean (C3),

518 wheat (C3) and millet (C4); for common beans (C3), the effect of weeds became positive with
 519 drought. Under warming, the impact of weeds remained negative for rice and decreased for
 520 barley, soybean, wheat and tomato becoming non-significant.

521



522

523 **Figure 7.** Effect size (ES) estimates of weeds, either under current environmental (ES1) or with
 524 environmental change (ES3), for studies of particular crop species. Credible intervals (95%) that
 525 do not include zero are considered statistically significant (indicated by an asterisk). Numbers
 526 indicate sample sizes. See Table S2 for parameter values.

525

526 Differences among crop species should be interpreted with caution due to the uneven taxonomic
 527 and geographical distribution of the studies (Fig. 1) and the small number of observations on the
 528 combined effects of weeds under environmental change for many crops. More than half of the
 529 observations used rice, soybean and tomato crops, while nine crop species were represented by
 530 a single observation (Fig. 7). We should also be aware that differences in weed composition and
 531 densities across observations might influence their impact (Vilà *et al* 2004, Zimdahl 2004).

532 Conclusions and the way forward

533 Understanding how global change will affect crop yield is critical for projecting future food
 534 production. For this reason, many studies have quantified the effects of two major factors
 535 affecting crop yields: climate change and weeds. However, most studies have examined these
 536 factors in isolation (Juroszek and Von Tiedemann 2013), leaving uncertainty about the validity of
 537 extrapolations (Ward *et al* 2014). Studies that simultaneously address the effects of
 538 environmental conditions related to climate change and weeds on crops are not common, and
 539 surprisingly, many have experimental design limitations that precluded their inclusion in meta-

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2
3 540 analysis (Gurevitch *et al* 1992). Many studies did not explore the single and interactive effects of
4 541 weeds and environmental change under the same experimental conditions or on the same crop
5 542 varieties. Other studies lacked control treatments, had no replication, or did not present variance
6 543 data. This information is often missing in agronomic studies of competition (Vilà *et al* 2004), which
7 544 limited the dataset of studies available for synthesis. To present a comprehensive dataset, we
8 545 included studies where the primary aim was not to test for the effect of climate change, but which
9 546 provided proxies (i.e., contrasting environmental differences) to test for the effect of
10 547 environmental change on crops with and without (or with low levels of) weeds. To better assess
11 548 how climate change affects weed constraints on crops, future research should implement
12 549 replicated well designed experiments with controls that provide full statistics and that explicitly
13 550 test realistic environmental changes in field conditions. Future studies should also evaluate the
14 551 effects of multiple environmental change components on crops with and without weeds (Peters *et*
15 552 *al* 2014).

16 553 Of all pests, weeds have the greatest potential to reduce worldwide crop yields (Oerke 2006).
17 554 Moreover, our meta-analysis indicates that the effects of weeds alone can be more detrimental on
18 555 crop yield than environmental change alone. Our results also suggest that weeds will reduce crop
19 556 yield under climate change by a similar magnitude to their effects under current climatic
20 557 conditions. Therefore, weed management will remain a critically important activity climatic
21 558 change. Weed management is facing major challenges such as the increasing rates of weed
22 559 dispersal through global trade and climate change, the environmental damage caused by weed
23 560 control, and weed resistance to herbicides (Liebman *et al* 2016). Because our results indicate that
24 561 under forecasted climate change, the negative effects of weeds will persist to similar magnitude,
25 562 we propose the following priority research areas: (i) comparing the effects of different weed
26 563 management practices (e.g. chemical vs. mechanical) to minimize crop yield losses and costs
27 564 under climate change (Peters *et al* 2014); (ii) focusing on rarely studied subsistence crops (e.g.
28 565 vegetables) that depend on manual labor for weed management and on farming systems that
29 566 cannot compensate for drought with irrigation (Altieri 2019); (iii) exploring differences among crop
30 567 varieties (e.g. weed-suppressive crop genotypes) in the impact of weeds and climate change
31 568 (Korres *et al* 2016), (iv) conducting research in regions where there are few studies, such as in
32 569 the southern hemisphere, especially on weed effects with warming and (v) exploring if there are
33 570 thresholds of environmental change that might cause non-additive effects with weeds.

34 571

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36 573

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581

582 Main data that support the findings of this study are included in Table S1.

583

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