



1 **Global meta-analysis of the relationship between soil organic matter and crop yields**

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15 **Abstract**

16 Resilient, productive soils are necessary to sustainably intensify agriculture to increase
17 yields while minimizing environmental harm. To conserve and regenerate productive
18 soils, the need to build and maintain soil organic matter (SOM) has received considerable
19 attention. Although SOM is considered key to soil health, its relationship with yield is
20 contested because of local-scale differences in soils, climate, and farming systems. There
21 is a need to quantify this relationship to set a general framework for how soil
22 management could potentially contribute to the goals of sustainable intensification. We
23 developed a quantitative model exploring how SOM relates to crop yield potential of
24 maize and wheat in light of co-varying factors of management, soil type, and climate. We
25 found that yields of these two crops are on average greater with higher concentrations of
26 SOC. However, yield increases level off at ~2% SOC. Nevertheless, approximately two
27 thirds of the world's cultivated maize and wheat lands currently have SOC contents of
28 less than 2%. Using this regression relationship developed from published empirical data,
29 we then estimated how an increase in SOC concentrations up to regionally-specific
30 targets could potentially help reduce reliance on nitrogen (N) fertilizer and help close
31 global yield gaps. Potential N fertilizer reductions associated with increasing SOC
32 amount to 7% and 5% of global N fertilizer inputs across maize and wheat fields,
33 respectively. Potential yield increases of $10 \pm 11\%$ (mean \pm SD) for maize and $23 \pm 37\%$ for
34 wheat amount to 32% of the projected yield gap for maize and 60% of that for wheat. Our
35 analysis provides a global-level prediction for relating SOC to crop yields. Further work
36 employing similar approaches to regional and local data, coupled with experimental work



- 37 to disentangle causative effects of SOC on yield and vice-versa, are needed to provide
- 38 practical prescriptions to incentivize soil management for sustainable intensification.



39 **1 Introduction**

40 The pressure to increase crop production has resulted in the expansion of land area
41 dedicated to agriculture and the intensification of cropland management through practices
42 such as irrigation and fertilization. These practices have led to degradation of land and
43 waters prompting “sustainable intensification” initiatives to increase yields on existing
44 farmland while decreasing the environmental impact of agriculture (Foley et al., 2011;
45 Godfray et al., 2010; Mueller et al., 2012). One sign of land degradation is the loss of soil
46 organic matter (SOM) (Reeves, 1997). Re-building SOM in agricultural lands holds the
47 promise of improving soil fertility, as SOM affects many properties of soils, including
48 their ability to retain water and nutrients, to provide structure promoting efficient
49 drainage and aeration, and to minimize loss of top-soil via erosion (Reeves et al., 1997;
50 Robertson et al., 2014). As such, managing SOM to ensure stable and long-lasting crop
51 productivity, and to decrease reliance on external inputs such as mineral fertilizers and
52 irrigation, has been identified as a critical component of sustainable intensification (Foley
53 et al., 2011). Yet the emphasis on soil management has remained qualitative, meaning
54 that the potential contribution of building SOM as a means to increase crop production
55 and minimize the environmental impact of agriculture has not yet been broadly quantified
56 (Adhikari and Hartemink, 2016; Chabbi et al., 2017; Hatfield et al., 2017).

57 A primary hurdle to managing SOM for sustainable intensification is the lack of
58 predictive, quantitative targets of SOM for specific agricultural and environmental
59 objectives (Herrick, 2000; NRC, 2010). While several studies show correlations between
60 SOM and yield (Culman et al., 2013; de Moraes Sa et al., 2014; Lucas and Weil, 2012;
61 Stine and Weil, 2002), it remains unclear how much yield could be expected to increase



62 per unit change in organic matter (Herrick, 2000; NRC, 2010). Establishing these
63 quantitative metrics is challenging because research shows increases (Bauer and Black,
64 1992), decreases (Bhardwaj et al., 2011), and no change (Hijbeek et al., 2017) in yields
65 with increased SOM. This lack of a general relationship is likely the result of a number of
66 interacting factors related to management, climate, and soil type that can confound the
67 SOM-yield relationship. This confusion has led some to claim that the amount of SOM is
68 unnecessary for crop yields, so long as there is sufficient N fertilizer (Hijbeek et al.,
69 2017; Loveland and Webb, 2003; Oelofse et al., 2015); whereas others highlight the need
70 to build SOM to increase crop yields while minimizing environmental harm (Lal, 2004).
71 The growing momentum to launch global scale initiatives to manage SOM (Banwart et
72 al., 2014; Lal, 2004; Minasny et al., 2017; Zomer et al., 2017) suggests the need to test
73 competing claims about the effects of SOM on these agricultural and environmental
74 outcomes.

75 One could critique the effort to establish a global-level understanding of the
76 SOM-yield relationship on the grounds that farm-level responses are necessarily
77 heterogeneous and poorly predicted by global assessments. Yet, global initiatives for
78 managing SOM could create policy environments that stimulate regional- and local-
79 prescriptions for SOM levels that inform practice (Chabbi et al., 2017; Minasny et al.,
80 2017; Zomer et al., 2017). Whereas it is difficult to disentangle the extent to which SOM-
81 yield relationships are driven by SOM effects on yield, as opposed to yield (i.e. higher
82 plant carbon inputs) effects on SOM, there is nevertheless experimental evidence
83 showing that building SOM positively affects yield (Bauer and Black, 1994; Majumder et
84 al., 2008; Oldfield et al., 2017). In addition, numerous soil properties that relate to soil



85 fertility, such as water holding capacity, respond positively to increasing SOM and in
86 turn are expected to increase yields (Williams et al., 2016). As such, correlative SOM-
87 yield relationships suggest the potential – but likely not the true – effect of SOM on yield.

88 We developed a quantitative model exploring how SOM relates to crop yield
89 potential in light of co-varying factors of management, soil type, and climate. The aim is
90 that this model can then be used to inform actionable and evaluable targets for soil
91 management as a central component of sustainable intensification efforts. We quantified
92 the relationship between SOM (measured as SOC, a common proxy for SOM) and yield
93 at a global level using data from published studies (Fig. 1). We focused our analyses on
94 wheat and maize, two common staple crops that (along with rice) constitute two-thirds of
95 the energy in human diets (Cassman, 1999). Along with SOC, we modeled the effects on
96 crop yields of several factors widely reported in yield studies: N input rate, irrigation, pH,
97 soil texture (% clay), aridity, crop type (i.e. wheat or maize), and latitude (as a proxy for
98 growing-season day length). The data informing our model came from empirical studies
99 that capture local scale variation in these variables, and hence we interpret our results in
100 light of the correlative nature of the database we assembled. Using the resulting multiple-
101 regression relationship, we then estimated how an increase in SOC concentrations up to
102 regionally-specific target thresholds might affect global yields. Our overarching aim was
103 then to estimate the potential extent to which restoring SOC in global agricultural lands
104 could help close global yield gaps and potentially help reduce reliance on – and the
105 negative effects of – N fertilizer.

106

107 **2 Results and Discussion**



108 **2.1 The relationship between SOC and yield**

109 At the global level and focusing specifically on the potential effect size of SOC on yield,
110 we found that the largest gains in yield occur between SOC concentrations of 0.1 to
111 2.0%. For instance, yields are 1.2 times higher at 1.0% SOC than 0.5% SOC (Fig. 2).
112 Gains in yield leveled off (i.e. the slope between yield and SOC is <0.25 , when
113 controlling for N input) at a concentration of approximately 2% SOC (Fig. 2). Two
114 percent SOC has previously been suggested as a critical threshold, with values below this
115 concentration threatening the structure, and ultimately, the ability of a soil to function
116 (Kemper and Koch, 1966). Importantly, the asymptotic relationship between SOC and
117 yield lends support to the idea that building SOC will increase yields – at least to a
118 certain extent – as opposed to simply being an outcome of higher yields. That is, higher
119 yields might be expected to lead to greater plant carbon inputs to soils, but given that 2%
120 SOC is well below the carbon saturation point for most soils (Castellano et al., 2015), the
121 asymptote supports a causal effect of SOC.

122 It has been suggested that there is no evidence for 2% SOC being a critical
123 threshold for productivity, as long as there is sufficient mineral fertilizer to support crop
124 production (Edmeades, 2003; Loveland and Webb, 2003; Oelofse et al., 2015). Such
125 conclusions deem the amount of SOM as substitutable by mineral fertilizers for crop
126 growth, but are inconsistent with the motivation for sustainable intensification to
127 minimize environmental harm caused by mineral fertilizers in relation to emissions of
128 greenhouse gases and eutrophication of waters (Vitousek et al., 2009). This logic also
129 does not account for the other co-benefits associated with building SOM in agricultural
130 lands such as reductions in nutrient run-off, drought resistance, and yield stability



131 (Robertson et al., 2014). Specific field and regional studies have shown a similar pattern
132 as that observed from our global analysis: there exists a positive relationship between
133 SOC and yield that starts to level off at ~2% SOC (Kravchenko and Bullock, 2000; Pan
134 et al., 2009; Zvomuya et al., 2008). Our analysis suggests that this relationship holds on
135 average at the global scale and when N fertilization is controlled for.

136 Ninety-one percent of the published studies used for our analysis were carried out
137 in fields with less than 2% SOC, with a mean of 1.1%. To see whether these observations
138 in SOC distribution reflected global patterns, we used globally gridded data on crop yield
139 and SOC (to a depth of 15 cm) (Hengl et al., 2014; Monfreda et al., 2008). We found that,
140 by both area and production, two thirds of maize and wheat cultivation takes place on
141 soils with less than 2% SOC (Fig. 3). Indeed, a recent analysis estimates that agricultural
142 land uses (including cropland and grazing) have resulted in a loss of 133 Pg of carbon
143 over the past 12,000 years of human land use (Sanderman et al., 2017). There appears to
144 be, therefore, significant opportunity to increase SOC on maize and wheat lands to
145 improve crop yields.

146

147 **2.2 The interaction between SOC and N fertilizer on yield**

148 One of the key goals of sustainable intensification is to reduce the environmental impacts
149 of agriculture (Foley et al., 2011; Mueller et al., 2012). Nitrogen fertilization, while a
150 boon to yields, can cause environmental damages, such as eutrophication of waters and
151 increased soil emissions of nitrous oxide, a potent greenhouse gas (Vitousek et al., 2009).
152 Using our regression model, we asked whether there might be target N fertilizer addition
153 rates that suggest the possibility of maximizing yield per unit N applied by building SOC



154 and reducing N inputs. We wanted to see if yields converge at higher levels of SOC,
155 suggesting that crops are achieving sufficient nutrients through SOM and excess mineral
156 N is not necessary. Our analysis suggests that SOC is not directly substitutable for
157 mineral fertilizer (Fig. 2); however, at lower rates of N input (≤ 50 kg N ha⁻¹), we found
158 that increasing SOC from 0.5 to 1.0% could potentially maintain current yields and
159 reduce fertilizer inputs by approximately half (50%). At higher rates of N input (≥ 200 kg
160 N ha⁻¹), an increase from 0.5 to 2.0% SOC could potentially reduce N inputs by up to
161 70% per hectare (Fig. 4). These results suggest that building SOM in agricultural lands
162 may supply enough plant available nutrients to sustain crop yields while drastically
163 cutting back on N fertilizer inputs.

164 There was an interaction between SOC and N input, where at higher SOC
165 concentrations N input had a greater impact on yield (Fig. 2, Table 1). This may be
166 because higher SOC improves soil structure and water holding properties, resulting in
167 improved crop growth at a given level of N input (Powlson et al., 2011). Additionally,
168 soils receiving more N may have greater SOC because N increases crop yields, which can
169 increase the return of plant residues into the soil and potentially build SOC (Powlson et
170 al., 2011). However, if the relationship was simply an effect of greater inputs building
171 SOC, we should not have seen an interaction between SOC and N on yields (because
172 SOC should then just have been additively related to yield). Whatever the specific
173 explanation, the SOC by N interaction we detect suggests that a combination of both
174 building SOM and using targeted N applications could lead to potential increases in yield
175 (Fig. 4). Practices such as cover-cropping represent a strategy that can both increase N
176 supply and build SOM through biological N fixation and the return of high quality



177 residues (narrow C:N ratios) to the soil (Drinkwater et al., 1998). The combination of
178 both SOM improvement and targeted fertilizer input will likely be especially important
179 for degraded soils, which require a suite of organic and inorganic nutrients to help build
180 SOM and improve crop yields (Palm et al., 1997).

181 Gains in yield from fertilizer input leveled off at about $200 \text{ kg N ha}^{-1} \text{ y}^{-1}$ (Fig. 4),
182 meaning that optimum yields appear achievable, at least on average, with this fertilizer
183 input level and an SOC target concentration of 2%. Using this target N input rate, we
184 explored potential fertilizer reductions on agricultural lands using more than 200 kg N ha^{-1}
185 y^{-1} . We found that for lands receiving more than $200 \text{ kg N ha}^{-1} \text{ y}^{-1}$, current yields could
186 be maintained, while decreasing global N fertilizer inputs by 7% for maize and 5% for
187 wheat. It has been estimated that 25 to 30% of fertilizer N is exported to streams and
188 rivers, resulting in eutrophication (Raymond et al., 2012). Targeted reductions in the
189 application of fertilizer N on the order of magnitude our analysis suggests could then
190 prevent the annual export of as much as 3.73 million tonnes of N into inland waters,
191 which amounts to 10% of mineral fertilizer applied to maize and wheat lands.

192

193 **2.3 Exploring potential reductions in global yield gaps of maize and wheat**

194 With a majority of cultivated lands containing less than 2% SOC and a growing
195 imperative to build, restore, and protect SOC in agricultural soils (NSTC, 2016; FAO,
196 2008; NRCS, 2012), we used global gridded data sets coupled with our regression model
197 (Table 1) to examine the potential gains in yield and production if opportunities to
198 increase SOC are realized (Table 2). Although our model identified 2% as a global target
199 for SOC, we created regionally-specific SOC targets given the fact that achieving 2%



200 SOC in some soils (e.g. those of drylands) may be unachievable due to inherent
201 constraints of physical soil properties and climate (Stockmann et al., 2015). We created
202 region-specific SOC targets for each of the 18 agro-ecological zones (AEZ) defined by
203 the Food and Agriculture Organization (Ramankutty et al., 2007). We identified a target
204 of 2% for 14 of the 18 AEZs, and targets ranging from 1 to 1.5% for the remaining four,
205 arid AEZs (see Methods). These targets were in line with recent quantitative assessments
206 based on similar climatic classifications (Stockmann et al., 2015). We then used global
207 data sets on N input rate (Mueller et al., 2012), SOC (Hengl et al., 2014), pH (Hengl et
208 al., 2014), texture (Hengl et al., 2014) and aridity (Zomer et al., 2008) to fill each term in
209 our regression model to explore how increasing SOC concentrations to the regional
210 targets could potentially affect global yields. We found that increasing SOC
211 concentrations to the defined targets has the potential capacity to increase average yields
212 on a per hectare basis by $10\pm 11\%$ (mean \pm SD) for maize and $23\pm 37\%$ for wheat. These
213 gains in yield translate to a 5% and 10% increase in the global annual tonnes produced of
214 maize and wheat, respectively (Table 2). These increases in production would close 32%
215 of the global yield gap for maize and 60% of the gap for wheat (Fig. 5a, b).

216 These yield gap results represent an exploration of potential “best case” impacts
217 of increasing SOC concentrations. We recognize there are inherent and logistical
218 challenges to building SOM in agricultural soils. For instance, soil characteristics such as
219 texture have a large effect on SOC content because sandier (rather than more clay rich)
220 soils have less surface area to stabilize SOC (Cotrufo et al., 2013), and so hold much less
221 water and nutrients than clay-rich soils (Johnston et al., 2009). Maintaining SOC contents
222 in sandy soils may require more frequent additions of organic amendments because these



223 soils do not have the surface area to retain nutrients, moisture, and to stabilize SOC
224 (Lehmann and Kleber, 2015). Furthermore, different regions and climate types face
225 different imperatives for building SOM. In the mid-western United States, for instance,
226 building SOM may be a good strategy to reduce fertilizer inputs and irrigation needs;
227 whereas in sub-Saharan Africa, building SOM may be critical for drought protection and
228 nutrient provision. Notably, high SOM values are not common in dryland environments
229 (for our dataset, mean SOC = 0.9% for dryland climates versus 1.4% SOC for mesic
230 soils), and building and maintaining SOM in arid zones is typically hindered by the lack
231 of organic matter to return to soils (Rasmussen et al., 1980). On a positive note, however,
232 our analysis suggests that increases in SOC in drylands, for example, from 0.5 to 0.8%,
233 could potentially increase yields by 10%, likely due to impacts on water retention.

234 Whereas we did use lower SOC targets (ranging from 1.0 to 1.5%) for the arid
235 AEZs in our analysis, the majority of data used for our analysis is from the more
236 temperate and tropical humid zones (Fig. S2). To bolster and/or refine SOC targets, our
237 correlative analysis needs to be supplemented with well-replicated experimental studies
238 incorporating different management strategies across multiple soil and climate types to
239 develop SOC-yield relationships that can be applied to the specific set of local farm
240 conditions. Further, these studies should ideally report data related to soil texture and
241 mineralogy, nutrient management, and paired SOC-yield observations with SOC taken to
242 meaningful depths, such as those that represent plant-rooting depth. These experimental
243 studies will help generate information that practitioners can use to inform management by
244 taking into account the inherent and logistical challenges to building SOM in agricultural
245 soils.



246

247 **3 Conclusions**

248 Despite uncertainties and calls for further research into how SOM affects agricultural
249 performance (Cassman, 1999; Herrick, 2000; Oldfield et al., 2015), policy for sustainable
250 intensification already widely supports the merits of increasing SOM in agricultural lands
251 (FAO, 2008; NRCS, 2012). The purported benefits include improved yields, increased
252 resilience, and decreased inputs of fertilizer and irrigation water. However, although
253 consensus exists around the importance of SOM to soil health, translating SOM policy to
254 practice is hindered by the lack of a predictive capacity for SOM target setting to inform
255 management efforts focused on yield and reducing fertilizer and irrigation (Chabbi et al.,
256 2017; Herrick, 2000; NRC, 2010). Our analysis helps establish a quantitative framework
257 for SOC targets that achieve measurable agricultural outcomes as part of sustainable
258 intensification efforts. It quantifies the potential effect size of SOC on yield while also
259 accounting for climate, soil, and management variables that influence crop yield. We find
260 that greater concentrations of SOC are associated with greater yields up to an SOC
261 concentration of 2%. With two thirds of global maize and wheat lands having SOC
262 concentrations of less than 2%, there seems significant opportunity to increase SOC to
263 reduce N inputs and potentially help close global yield gaps.

264

265 **4 Methods**

266 Our approach consisted of a two-stage process. In the first stage, we assembled published
267 empirical data from studies that reported both SOC and yield data for maize and wheat.
268 From this meta-dataset, we then quantified how both SOC concentrations and N input



269 rates are related to yields, in the context of spatial variation in climatic, management, and
270 soil co-variables. In the second stage, we used globally-gridded data sets to extract values
271 for the factors we investigated in the first stage for global lands where maize and wheat is
272 produced. Using the regression relationship developed from the published empirical data
273 compiled under the first stage, we then estimated how an increase in SOC concentrations
274 up to target thresholds we identified (ranging from 1 to 2% depending on agro-ecological
275 zoning) affected global yield potentials. Finally, we used an N input threshold identified
276 through our regression analysis ($200 \text{ kg N ha}^{-1} \text{ yr}^{-1}$) to calculate potential N reductions on
277 global maize and wheat lands.

278

279 **4.1 Data collection**

280 In the first stage of our approach, we searched the database Web of Science (Thomson
281 Reuters) in January 2016 and again in October 2016 using the following topic search
282 terms: “soil organic matter” OR “soil organic carbon” OR “soil carbon” OR “soil c”
283 AND “yield” OR “crop yield” OR “productivity” AND “agricult*.” We restricted the
284 initial search to articles published in English between 1980 through December 2015, and
285 excluded conference proceedings; the second search captured articles published in 2016.
286 The initial search resulted in 1,384 articles and the second 169 articles (Fig. S1). For each
287 citation, we reviewed titles and abstracts to select articles that met the following criteria:
288 experimental field studies whose abstract included information on yield and SOC for
289 systems growing wheat and/or maize. This initial screening resulted in 523 records for
290 which we assessed the full text. We assessed these records for eligibility based on
291 inclusion of data on crop yield, SOC, and N fertilizer rates for each observation. For



292 inclusion within our analysis, it was essential that studies reported paired SOC and yield
293 data. Furthermore, we required SOC concentrations (as opposed to stocks). Studies did
294 not meet our criteria for inclusion if they reported SOC stocks with no corresponding data
295 on bulk density to convert into concentrations; and also if they reported baseline SOC
296 concentrations as opposed to experimental SOC concentrations that we could pair with
297 yield data. In addition to our literature search, we also contacted authors to see if they
298 were willing to include raw data within our database. This resulted in three datasets
299 (Adiku et al., 2009; Birkhofer et al., 2008; Kautz et al., 2010). Finally, we consulted the
300 recently published database by the Swedish Board of Agriculture that is a key repository
301 of peer reviewed literature focusing specifically on studies (735 in total) related to the
302 effects of agricultural management on soil organic carbon (Söderström et al., 2014). We
303 explored this database to find studies from regions that were under-represented within our
304 literature search (e.g. the southern hemisphere). This resulted in a search of 55 studies to
305 see if they met our criteria for inclusion. We scanned each paper to see if they included
306 SOC data paired with matching yield data. From these papers, we extracted data from 12
307 studies, which resulted in an additional 52 data points. We encountered limitations
308 similar to our initial search: Namely, SOC and yield data were not paired, studies
309 included only baseline SOC concentrations, or SOC stocks were reported without any
310 corresponding bulk density data to convert into concentrations. Overall, our data set
311 included 840 individual observations from 90 articles covering sites across the globe
312 (Fig. 1) (Adiku et al., 2009; Agegnehu et al., 2016; Albizua et al., 2015; Alijani et al.,
313 2012; Araya et al., 2012; Atreya et al., 2006; Bai et al., 2009; Bedada et al., 2014;
314 Bhardwaj et al., 2011; Bhattacharyya et al., 2015; Birkhofer et al., 2008; Boddey et al.,



315 2010; Boulal et al., 2012; Bremer et al., 1994; Calegari et al., 2008; Campbell et al.,
316 2007; Castellanos-Navarrete et al., 2012; Celik et al., 2010; Chen et al., 2015; Chirinda et
317 al., 2010; Cid et al., 2014; Costa et al., 2010; D'Hose et al., 2014; Datta et al., 2010;
318 DeMaria et al., 1999; Diacono et al., 2012; Grandy et al., 2006; Guo et al., 2012; 2009;
319 He et al., 2011; Hossain et al., 2016; Hu et al., 2015; 2014; Kaihura et al., 1999;
320 Karbozova Saljnikov et al., 2004; Kautz et al., 2010; Kazemeini et al., 2014; Kucharik et
321 al., 2001; Larsen et al., 2014; Lebbink et al., 1994; Leogrande et al., 2016; Li et al., 2015;
322 Liu et al., 2014a; 2016; 2014b; 2014c; López-Garrido et al., 2014; Lu et al., 2016; Ma et
323 al., 2012; 2016; Madejón et al., 2001; Mandal et al., 2013; Masto et al., 2007; Mikanová
324 et al., 2012; Mishra et al., 2015; Mupangwa et al., 2013; N'Dayegamiye, 2006; Niu et al.,
325 2011; Njoku and Mbah, 2012; Paul et al., 2013; Qin et al., 2015; Quiroga et al., 2009;
326 Sadeghi and Bahrani, 2009; Saikia et al., 2015; Scalise et al., 2015; Seremesic et al.,
327 2011; Singh and Dwivedi, 2006; Singh et al., 2016; Sisti et al., 2004; Soldevilla-Martinez
328 et al., 2013; Spargo et al., 2011; Šimon et al., 2015; Tejada et al., 2016; Tiecher et al.,
329 2012; van Groenigen et al., 2011; Vieira et al., 2007; 2009; Wang et al., 2015; 2014a;
330 2014b; Wortman et al., 2012; Wu et al., 2015; Yang et al., 2013; 2015a; 2015b; Yeboah
331 et al., 2016; Zhang et al., 2015; 2009; 2016; Zhao et al., 2016). Where necessary, we
332 extracted data from manuscript figures using GraphClick Software (v. 3.0.3,
333 <http://www.arizona-software.ch/graphclick/>).

334 Studies that presented individual data points recorded over multiple years were
335 included as well as studies that averaged both yield and SOC data over multiple years. To
336 avoid over-representation of studies that included data points recorded for both yield and



337 SOC over multiple years (>10 y), we took observations from the beginning, middle, and
338 last year of the study.

339

340 **4.2 Data compilation**

341 For each extracted observation, we compiled the following information: latitude,
342 longitude, year of data collection, crop type, yield, SOC or SOM, depth of SOC or SOM
343 measurement, N fertilization rate, P fertilization rate, soil pH, texture, and whether or not
344 crops were irrigated. We used SOC (as opposed to SOM) for our analysis given that SOC
345 is a common proxy for SOM. Carbon, as an element that is easily identified and
346 measured within soil, is thought to comprise ~50-60% of SOM and is commonly reported
347 in the literature (Pribyl, 2010). When SOM was reported, we converted it to SOC by
348 dividing the value by 1.724 (Cambardella et al., 2001). Different studies reported SOC
349 concentrations to different depths, which ranged from 0-5 cm to 0-30 cm, with the
350 majority of studies reporting SOC to 0-20 cm. When studies reported SOC to multiple
351 depths, we averaged SOC values across depths to 30 cm. If no information on irrigation
352 was provided, we scored the observation as rain-fed. Soil texture and pH were not
353 reported for every study; 79% of included studies reported pH, and so we used the
354 study's latitude and longitude to extract these data using ISRIC SoilGrids (Hengl et al.,
355 2014) to fill in the missing pH values. Texture was reported for about half (49%) of
356 included studies, and so we used coordinates to pull these data from SoilGrids as well
357 (Hengl et al., 2014). We also used latitude and longitude to obtain an "aridity index"
358 through the CGIAR-CSI database (Zomer et al., 2008).

359



360 **4.3 Data analysis**

361 We used a linear mixed model (LMM) to analyze the observations we extracted from the
362 literature. Our model included SOC, N fertilizer rate, crop type (maize or wheat, coded as
363 a binary variable), irrigation (coded as a binary variable), aridity index, latitude, pH, and
364 texture (% clay) as fixed effects. To account for any spatial and temporal correlation
365 among the studies, we nested year within study as random effects (Bolker et al., 2009).
366 The LMMs were fit with a Gaussian error distribution in the “lme4” package for the “R”
367 statistical program (version 3.3.1), using the “lmer” function. The first stage of our data
368 analysis was to test the data distributions. We removed data points with N fertilization
369 rates $>600 \text{ kg N ha}^{-1}$ (4 data points) and yields $>18 \text{ t ha}^{-1}$ (2 data points) since these
370 represented outliers for our dataset (being beyond 3 times the inter-quartile range of the
371 meta-dataset) and are not representative of on-farm management practices or outcomes.
372 Our final model was based on 834 observations across 90 studies. We added quadratic
373 terms for both SOC and N input rate since these variables exhibited a nonlinear
374 relationship with yield. The square-root of the variance inflation factors (vif) was <2 for
375 all factors when included as main effects, indicating that collinearity was low among all
376 variables. As would be expected, there was a correlation between SOC and its quadratic
377 term and N input rate and its quadratic term. We calculated the r^2 values for our model
378 following Nakagawa and Schielzeth (2013) to retain the random effects structure. The r^2
379 of our model was 83% for the full model, with the fixed effects explaining 42% of
380 observed variance within our data set.

381 We based the choice of factors for inclusion in our model on the approach of
382 Hobbs et al. (2006), by only investigating factors where biological mechanism as to their



383 influence on yield is firmly established and where we were interested in their effect sizes
384 relative to one another. Also following Hobbs et al. (2006), we did not carry out model
385 selection. Operationally, there is substantial subjectivity and lack of agreement in model
386 selection approaches, with different decisions leading to markedly different conclusions
387 as to the influence of different factors. Instead, coefficients are generally most robust
388 when all terms are retained in a model, assuming that inclusion of each is biologically
389 justified.

390 To examine the effects sizes of the factors on yield, we took two approaches.
391 First, we compared the size of the standardized coefficients, where standardizing
392 involved subtracting the mean of the factor from each observed value and dividing by
393 two standard deviations (Gelman, 2008). Dividing by two standard deviations is useful
394 when binary predictors are included within regression models (in our case, crop type and
395 irrigation are coded as a binary predictors). This way, continuous and binary variables all
396 have a mean of 0 and a standard deviation of 0.5 (Gelman, 2008). This accounts for the
397 fact that the factors were measured on different unit scales (Table 1). Second, we
398 examined the influence of changing SOC concentration or N fertilization rates on yield.
399 To do this, we used the regression relationship derived from our statistical model, held all
400 other factors at a constant value (e.g. the mean of all observations for that factor), and
401 systematically varied SOC or N fertilization across the range of values we extracted from
402 the literature. For SOC, this meant varying SOC values from 0.1 to 3.5% to estimate
403 changing yield of rain-fed maize or wheat as SOC concentrations were increased (Fig. 2).
404 For N fertilization, we varied N input rates from 0 to 300 kg N ha⁻¹ for rain-fed maize or
405 wheat at different SOC concentrations (Fig. 4). When these factor-yield relationships



406 were plotted, we identified threshold values where yield became minimally responsive to
407 SOC or N fertilization as the point where the slope of the relationship became <0.25 (for
408 SOC) and < 0.002 (for N fertilization).

409

410 **4.4 Global extrapolations**

411 We used the regression relationship developed in the first stage of our approach to predict
412 how building SOC concentrations would potentially affect global crop yield averages. To
413 obtain values for each of the factors in our regression model at a global scale, we used
414 globally gridded data products. Global SOC, pH, and texture data were taken from ISRIC
415 SoilGrids (Hengl et al., 2014) at a 10-km grid cell resolution to match the spatial grain for
416 maize and wheat yields and N fertilization data, which we obtained from the EarthStat
417 product (Monfreda et al., 2008; Mueller et al., 2012). SoilGrids has multiple layers for
418 SOC concentrations, and we used the 0-15 cm layer as the average depth to which SOC
419 was reported for our dataset was 0-20 cm. The aridity index was obtained from the
420 CGIAR-CSI database (Zomer et al., 2008). We used the resulting global dataset to
421 explore the potential impact of increasing SOC (up to regionally identified threshold
422 levels ranging from 1 to 2%) on yield for lands across the globe where maize and wheat
423 are produced.

424 To establish regionally appropriate SOC targets, we classified maize and wheat
425 producing areas by their agro-ecological zones (AEZ). The Food and Agricultural
426 Organization have 18 zones defined on the basis of combinations of soil, landform, and
427 climatic characteristics (Ramankutty et al., 2007). For each AEZ, we examined the
428 distribution of SOC in areas classified as naturally vegetated (e.g. not in urban or



429 agricultural land uses). We did this by stacking two GIS raster layers of SOC (SoilGrids)
430 and land use (Friedl et al., 2010), excluding agricultural and urban land use
431 classifications. We then extracted SOC data for each AEZ using a shape file outlining the
432 geographical extent of each AEZ (Ramankutty et al., 2007). Examining the distribution
433 of SOC across each AEZ, we identified targets based on the mean SOC value within each
434 zone. All but four zones had means greater than 2% SOC, so we set target values for
435 those zones at 2%. Mean SOC concentrations were lower for the more arid zones and so
436 we set those targets to 1% for AEZ 1 and 1.5% for AEZ zones 2, 3, and 7. Recent
437 analysis of global SOC concentrations across globally defined Ecoregions shows mean
438 values of SOC at or greater than 2% for all regions except land classified as desert and
439 xeric shrubland (Stockmann et al., 2015).

440 Prior to our global extrapolations, we performed a suite of data checks. We
441 wanted to ensure that global yields predicted using our regression model were
442 comparable to those from EarthStat. These checks helped validate the strength of our
443 extrapolations. Firstly, we explored the range of variation in variables from experimental
444 data used to generate our model as well as the range of global variation in variables we
445 project across. The range of our regressors encompasses the range of global variation,
446 except for aridity, in which case 4.6% percent of our projections fall in grids that have
447 axis conditions outside of our range of measurements. These values fall in extremely arid
448 systems, with aridity values of less than 0.1. In these extremely arid zones, we do make a
449 point to use much lower target SOC values, recognizing that achieving 2% SOC in these
450 very arid areas is not very likely. Secondly, using our regression model to predict global
451 yields for both maize and wheat (separately), we first removed all values from the



452 analysis that had predicted yields of less than 0 because negative yields are not possible.
453 This amounted to 0.004% of the total predictions for maize and 0.15% for wheat. For
454 clarification, we refer to predictions from our regression model as “predicted” or “model-
455 predicted.” We then calculated the proportional difference between model-predicted and
456 globally gridded yield data from EarthStat. We dropped all cells for which the
457 proportional difference between predicted and gridded data was >3-times. This threshold
458 represents the mean \pm half of the standard deviation for the distribution of the
459 proportional difference between predicted and EarthStat yield data. This amounted to
460 14% of cells for maize and 7% for wheat. The mean proportional difference between
461 predicted and gridded data was 0.85 ± 0.91 for maize (Fig. S3b) and 0.45 ± 0.87 for wheat
462 (Fig. S4b). The correlation between predicted and gridded data was $r=0.73$ for maize
463 (Fig. S3c) and $r=0.38$ for wheat (Fig. S4c). We also visualized overlap in the distribution
464 of model-predicted and gridded data. Model-predicted maize yield had a global mean of
465 $4.66 \pm 1.84 \text{ t ha}^{-1}$ and EarthStat had a global mean of $3.34 \pm 2.62 \text{ t ha}^{-1}$ (Fig. S3a). Model-
466 predicted wheat yield had a global mean of $3.18 \pm 1.66 \text{ t ha}^{-1}$ and EarthStat had a global
467 mean of $2.43 \pm 1.58 \text{ t ha}^{-1}$ (Fig. S4a).

468 We also compared the distribution of EarthStat yield data with observed yield
469 data from the studies included in our analysis. We found that correlation (r values)
470 between the gridded and collected data was 0.56 for maize and 0.39 for wheat. Average
471 observed maize yield was $5.61 \pm 3.32 \text{ t ha}^{-1}$ and wheat yield was $4.02 \pm 2.11 \text{ t ha}^{-1}$
472 (mean \pm SD). EarthStat maize yield, again, was $3.34 \pm 2.62 \text{ t ha}^{-1}$ and wheat yield was
473 $2.43 \pm 1.58 \text{ t ha}^{-1}$. These differences between predicted and EarthStat yield averages are
474 likely due to the fact that EarthStat data is based on regional census data, incorporating



475 much more variability in terms of management practices and skill than experimental field
476 studies.

477 After the data checks, we then used our model to extrapolate global yield
478 potentials of maize and wheat given increases in SOC. We masked EarthStat production
479 and cultivated area data layers for maize and wheat for cells that had SoilGrids SOC
480 concentrations of >2%. We compared the subsetted data (i.e. cultivated lands with < 2%
481 SOC) with the original data layers to determine the fraction of global maize and wheat
482 production and cropland that is on soils with less than 2% SOC. We used this subsetted
483 data along with our regression model to predict yields at current SOC levels. As stated
484 above, we used EarthStat, ISRIC SoilGrids, and CGIAR-CSI data layers to fill in the
485 values for each of the factors in our regression model. This new data layer was used as a
486 baseline with which to compare to potential gains in yield with an increase to SOC target
487 values. This created a second data layer with model-predicted yields given an increase in
488 SOC. We calculated the percentage increase in yield between these two layers (the
489 baseline and the improved-SOC layer) and multiplied this by EarthStat yield and
490 production data to determine potential gains in maize and wheat yields and production
491 (Table 2). We then used EarthStat yield gap data to see how such an increase in SOC
492 would reduce projected yields gaps. Using the new yield data layer (with yields at SOC
493 target values), we calculated the proportion of EarthStat yield gaps that was reduced for
494 both maize and wheat.

495 Finally, we used data on global N use (EarthStat) to explore potential reductions
496 in fertilizer use for both maize and wheat, separately. We used a value of 200 kg N ha⁻¹ y⁻¹
497 as our N input threshold, as this is the value from our regression model at which gains



498 in yields level off. We created a new data layer for those areas that have N input rates
499 greater than 200 kg N ha⁻¹ y⁻¹. We then calculated the potential N reductions, in tonnes,
500 by multiplying this new data layer by EarthStat cultivated maize and wheat lands,
501 separately. Finally, we divided the potential reduction in N input (in tonnes) by total N
502 input (in tonnes) as provided through the EarthStat data product.

503

504 **Data availability**

505 The dataset generated and analyzed during the current study is available through the
506 KNB repository: <https://doi.org/10.5063/F1RV0KWK>

507

508 **Author contributions**

509 EEO, MAB, and SAW conceived of the study. EEO and SAW performed data analysis.
510 EEO wrote the first draft of the manuscript. All authors contributed to data interpretation
511 and paper writing.

512

513 **Competing Interests**

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

515

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1073 **Table 1. Modeled regression coefficients with standard errors, standardized**
 1074 **coefficients, and *P* values for our regression model.**

Variable	Unstandardized Coefficients	Standardized Coefficients	<i>P</i> value
Intercept	-1.62 ± 1.71	5.59 ± 0.18	0.34
SOC	1.78 ± 0.59	1.44 ± 0.30	0.003
SOC ²	-0.46 ± 0.18	-0.66 ± 0.27	0.013
N input	0.018 ± 0.0014	2.71 ± 0.15	< 0.00001
N input ²	-0.000039 ± 0.0000036	-1.64 ± 0.15	< 0.00001
Irrigation	0.77 ± 0.35	0.77 ± 0.34	0.027
pH	0.053 ± 0.18	0.12 ± 0.42	0.77
Aridity	0.16 ± 0.51	0.12 ± 0.41	0.76
Crop Type	1.55 ± 0.15	1.54 ± 0.15	< 0.00001
Clay (%)	0.013 ± 0.014	0.29 ± 0.31	0.36
Latitude	0.055 ± 0.016	1.40 ± 0.41	0.00077
SOC*N input	0.0039 ± 0.00099	0.96 ± 0.25	0.00010

1075 The output of our linear mixed effect model (n=834). The full model explained 83% of
 1076 observed variability within the data set with fixed effects (included in the table)
 1077 accounting for 42% of the variability. Standardized coefficients allow for direct
 1078 comparison of the relative effect size of each modeled variable despite different scales on
 1079 which the variables are measured. For example, crop type's effect on yield is two-times
 1080 greater than that of irrigation. Crop type was coded as a binary variable with 0 for wheat
 1081 and 1 for maize. Irrigation was also coded as a binary variable with 0 for no irrigation and
 1082 1 for irrigation.

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1088 **Table 2. Scenarios for increases in yield and reductions in N input with an increase**
 1089 **in SOC concentration to target values.**

Scenario	Crop	Global Yield Average (t ha ⁻¹)	Increase in production (Mt)	Nitrogen input (Mt N ha ⁻¹)
Current condition	Maize	3.34 ± 2.62	NA	17.24
	Wheat	2.43 ± 1.58	NA	33.07
Increase SOC to target concentrations	Maize	3.93 ± 3.08	29.96	15.96
	Wheat	3.17 ± 2.06	55.41	32.04

1090 Values (mean±SD) represent current EarthStat yields and projected gains in yield and
 1091 production resulting from an increase in SOC concentration to target values for each
 1092 agro-ecological zone (targets ranged from 1.0 to 2.0%). We used our regression model to
 1093 determine potential gains in EarthStat yield and reductions in EarthStat N input. Global
 1094 yield averages represent tonnes produced per unit land area, whereas production
 1095 represents tonnes of maize and wheat produced globally.

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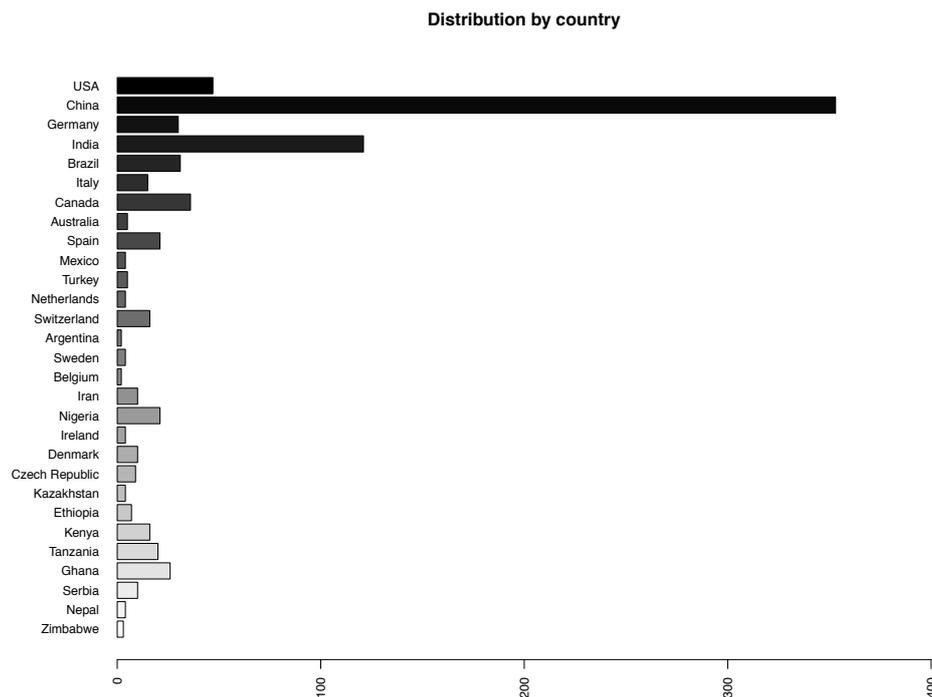
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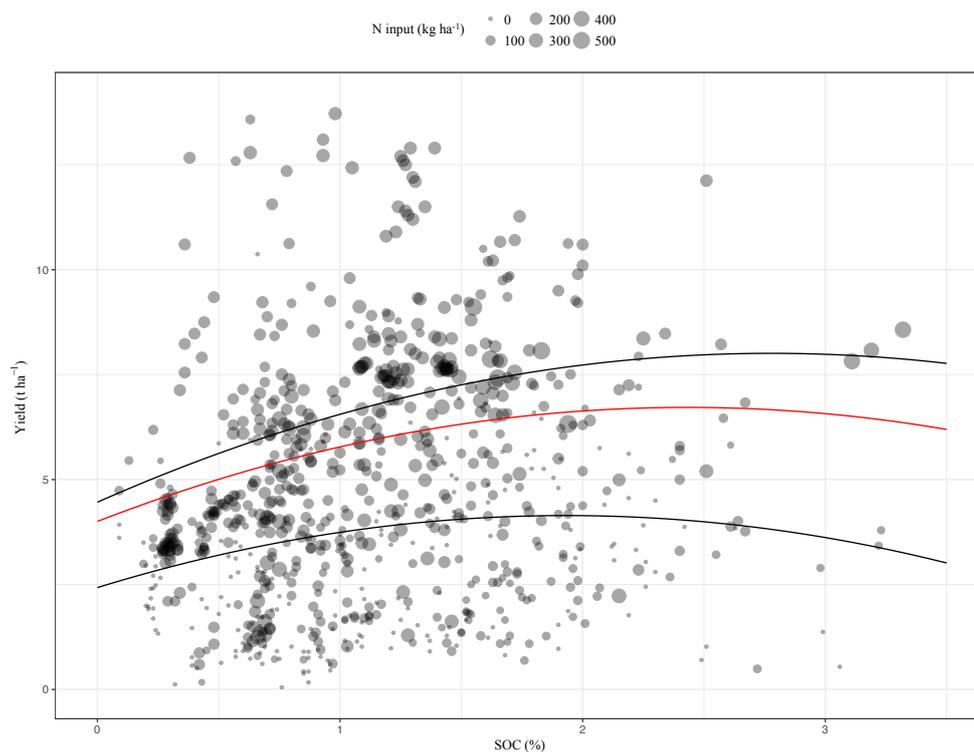
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1109 **Figure 1: Distribution of data points by country.** Countries are ordered by Gross
1110 Domestic Product (GDP) in order from largest (top) to smallest (bottom). The data set
1111 used for this study contains a total of 840 individual observations from 29 different
1112 countries.



1113

1114 **Figure 2: Relationship between SOC and yield of maize and wheat for published**

1115 **studies.** The regression lines are modeled yields (i.e. effect sizes) for rain-fed maize

1116 using observed means of our meta-dataset for aridity, pH, texture, and latitude at different

1117 N input rates. We varied SOC (x-axis) across the range of values extracted from the

1118 literature. The red line represents the mean N input rate ($118 \text{ kg N ha}^{-1} \text{ y}^{-1}$) across all

1119 studies, with the bottom line representing 0 inputs of N and the top line representing 200

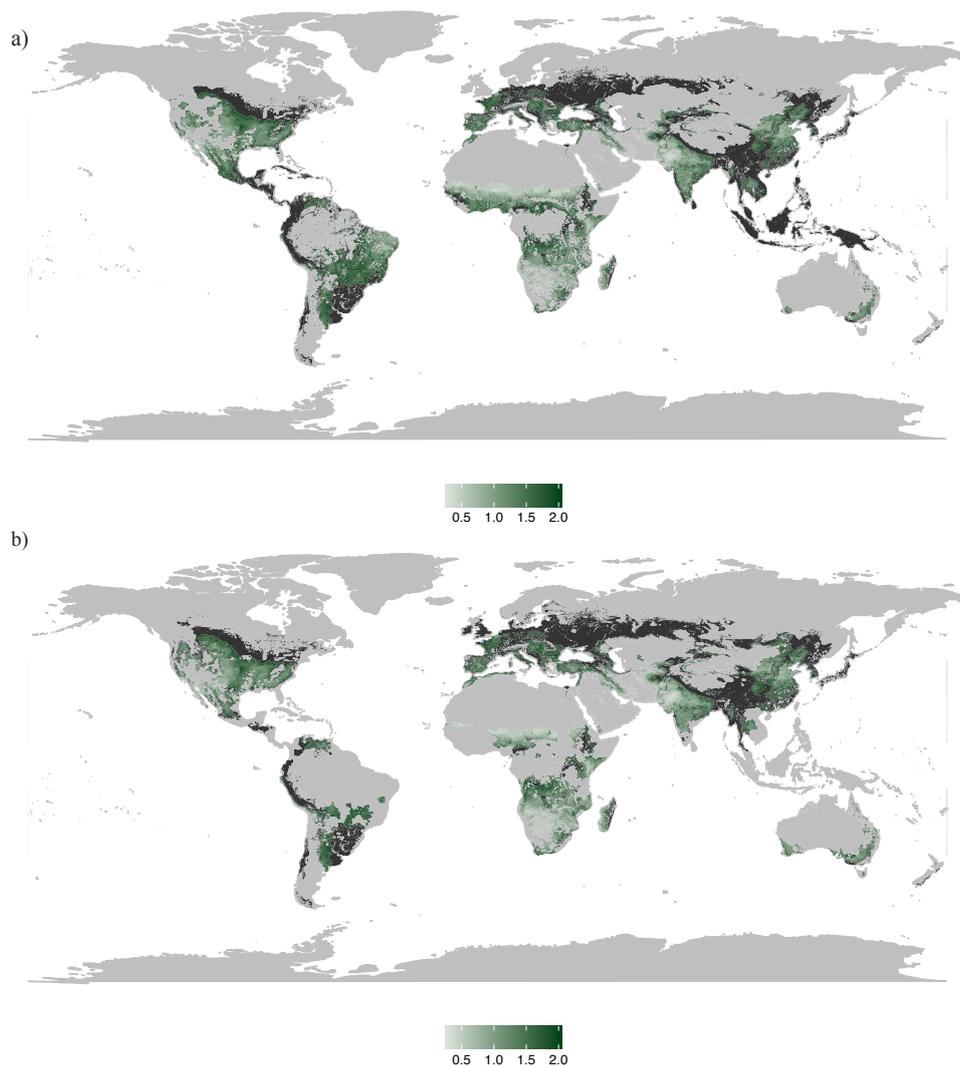
1120 $\text{kg N ha}^{-1} \text{ y}^{-1}$. For the raw data points, N input is mapped as a continuous variable across

1121 its range from 0 (smallest circles) to $500 \text{ kg N ha}^{-1} \text{ y}^{-1}$ (largest circles). Note that the

1122 observed scatter of the individual observations is an outcome of the fact yield is

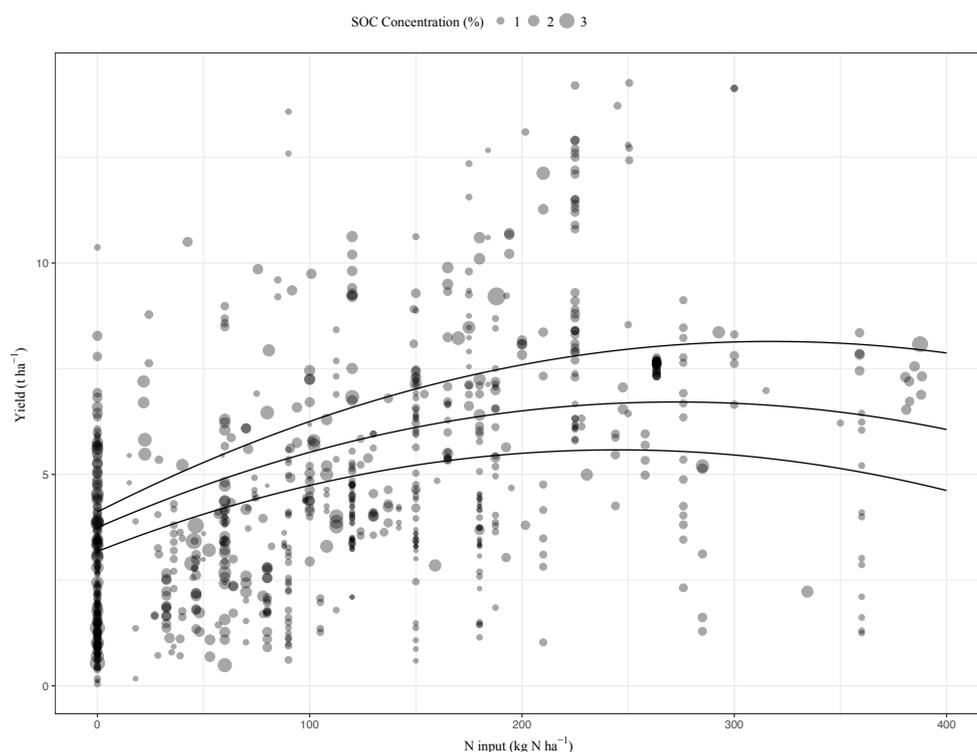
1123 controlled by multiple factors (Table 1), and therefore the regression lines isolate just the

1124 potential effect of SOC with all other factors held constant.



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1126 **Figure 3: Global maize and wheat lands with less than 2% SOC.** Cultivated (a) maize
1127 lands and (b) wheat lands on soils with SOC contents less than 2%. Approximately two
1128 thirds of all maize (61%) and of all wheat (64%) producing areas are on soils with less
1129 than 2% SOC. Black areas on the maps are cultivated maize and wheat lands that have
1130 concentrations over 2% SOC. Yield data is taken from EarthStat and SOC data is taken
1131 from ISRIC-SoilGrids.



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1133 **Figure 4: Potential reductions in nitrogen fertilizer with an increase in SOC**

1134 **concentration.** The lines on the graph represent varying SOC concentrations, 2.0%,

1135 1.0%, and 0.5% SOC from top to bottom. These lines are plotted on top of the

1136 observations from our data set with SOC mapped as a continuous variable across its

1137 range from 0.1% (smallest circles) to 3.0% (largest circles). Our model shows that

1138 keeping yield constant by increasing SOC contents allows for potentially significant

1139 reductions in N input (e.g. the same yield is achievable with 0 N input and 2% SOC, as

1140 with 65 kg N ha⁻¹ y⁻¹ and 0.5% SOC). Recognizing that the 0 N input values may

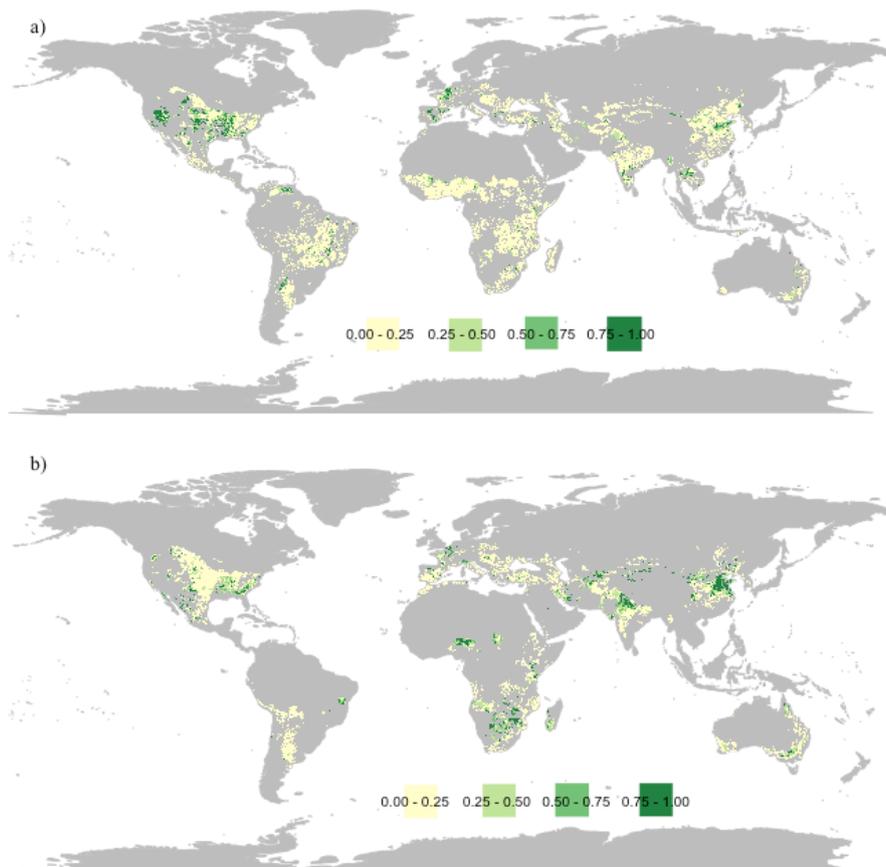
1141 influence the modeled relationship, we analyzed data excluding these values. The

1142 qualitative patterns remain the same if the 0 N input values are excluded from the

1143 analysis; and while the absolute quantitative patterns shift slightly, the general trends



1144 remain intact.



1145

1146 **Figure 5: Percentage closure of yield gap for (a) maize and (b) wheat given an**
1147 **increase in SOC concentration to target values for each AEZ (ranging from 1-2%).**

1148 Modeled gains come from our regression relationship between SOC and yield and
1149 applying it to EarthStat yield gap data. Doing so determines the potential increase in yield
1150 and therefore projected reductions in yield gaps for maize and wheat.

1151