1	Higher yields and lower methane emissions with new rice cultivars
2	Running Head: Reducing methane emissions from rice agriculture
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- 35 **Type of paper**: Primary research paper

Higher yields and lower methane emissions with new rice cultivars

37

38 Abstract

39 Breeding high-yielding rice cultivars through increasing biomass is a key strategy to meet 40 rising global food demands. Yet, increasing rice growth can stimulate methane (CH₄) emissions, exacerbating global climate change, as rice cultivation is a major source of this 41 42 powerful greenhouse gas. Here, we show in a series of experiments that high-yielding rice 43 cultivars actually reduce CH₄ emissions from typical paddy soils. Averaged across 33 rice 44 cultivars, a biomass increase of 10% resulted in a 10.3% decrease in CH₄ emissions in a soil 45 with a high carbon (C) content. Compared to a low-yielding cultivar, a high-yielding cultivar 46 significantly increased root porosity and the abundance of methane-consuming 47 microorganisms, suggesting that the larger and more porous root systems of high-yielding 48 cultivars facilitated CH₄ oxidation by promoting O₂ transport to soils. Our results were further 49 supported by a meta-analysis, showing that high-yielding rice cultivars strongly decrease CH₄ 50 emissions from paddy soils with high organic C contents. Based on our results, increasing rice biomass by 10% could reduce annual CH₄ emissions from Chinese rice agriculture by 7.1%. 51 52 Our findings suggest that modern rice breeding strategies for high-yielding cultivars can 53 substantially mitigate paddy CH₄ emission in China and other rice growing regions.

54

55 Introduction

Rice (*Oryza sativa* L.) is a staple food for more than half of the people in the world, and global demand for rice is projected to increase from 644 million tons in 2007 to a projected 827 million tons in 2050 (Alexandratos and Bruinsma, 2012). However, rice production is a major source of the potent greenhouse gas methane (CH₄); about 11% of anthropogenic CH₄ emissions come from rice paddies (IPCC, 2013), and among the major cereals, rice has the highest global warming potential (GWP) due to high CH₄ emissions (Linquist *et al.*, 2012). Therefore, sustainable intensification of rice cropping systems requires increasing yields while
reducing CH₄ emissions (Chen *et al.*, 2014; Linguist *et al.*, 2012; van Groenigen *et al.*, 2013).

Global rice production can be increased through improving yield potential of 64 65 rice cultivars; the introduction of high-yielding rice cultivars accounts for almost 50% of the 66 recent yield growth in developing countries (Evenson and Gollin, 2003; Peng et al., 2008). 67 Until the beginning of this century, breeding strategies to improve rice yield were mainly focused on increasing harvest index (Hay, 1995; Richards, 2000). This approach may lower 68 69 CH₄ emissions, as an increase in harvest index with constant plant biomass can decrease the 70 production of root exudates that fuel CH₄ production (van der Gon et al., 2002; Su et al., 71 2015). However, the current harvest index of high-yielding rice cultivars is about 0.55, 72 approaching the theoretical upper limit of 0.65 (Hay et al., 1995; Peng et al., 2008). 73 Therefore, more recent breeding strategies for increasing yields focus on enhancing biomass 74 while maintaining the current harvest index (Richards, 2000; Peng et al., 2008; Cheng et al., 75 2007; Yuan, 2015).

76 These latter breeding strategies could stimulate CH₄ emissions, because recent 77 photosynthate of rice plants can be a major substrate for CH₄ production: with higher biomass 78 production, more substrate could fuel higher CH₄ emission rates (Huang *et al.*, 1997; 79 Watanabe et al., 1999). The microorganisms that produce CH₄, methanogenic archaea, also 80 use substrates that are derived from native soil organic carbon (C) (Watanabe et al., 1999; 81 Conrad, 2007), suggesting that the effect of rice cultivars on CH₄ emissions depends on soil C 82 content. Thus, to study the effects of high-yielding cultivars on CH₄ emissions and their 83 possible interaction with soil C availability, we conducted three independent but 84 complementary experiments. 1) We used 33 rice cultivars to quantify the relationship between plant production and CH₄ emission in two otherwise similar paddy soils with different labile 85 86 soil C contents, 2) We determined the effect of a high-yielding cultivar on CH₄ emissions in a 87 realistic field setting, and 3) Using the same soils and cultivars as in experiment 2, we grew

- rice in microcosms and measured CH₄ emissions with and without wheat straw incorporation.
- 89 Finally, to test the generality of our findings, we conducted a meta-analysis of studies that

90 quantified the effect of high-yielding rice cultivars on CH₄ emissions.

91

92 Methods

93 Experiment 1

94 In this pot experiment, we quantified the relationship between plant production and CH₄ emission in two otherwise similar paddy soils with different labile soil C contents. The 95 96 experiment was conducted under open field conditions at Pailou experimental station, Nanjing 97 Agricultural University, Nanjing City (118.8° E, 32.1° N), China. Thirty-three rice cultivars 98 approved and released since 2001 in China (Table 1) were tested. Both soils in this 99 experiment were collected from the plow layer of the paddy fields at Jiangpu Farm of Nanjing 100 Agricultural University, Nanjing City. The low C soil was stored outdoors for three years 101 before being used; soil labile C content was low because most of the plant residues and other 102 labile C in the soil had been oxidized or mineralized. The soil with a high soil labile C content 103 was collected five days before the experiment. Soil labile C content was measured by the 104 KMnO₄ oxidation method (Blair et al., 1995). Soil properties are reported in Table 1. 105 Plastic pots (height, 25 cm; diameter, 24 cm) were filled with 7.0 kg of soil that was 106 sieved (6 mm mesh size) to remove stones. Fifteen pots were prepared for each cultivar with 107 each soil. Three pots were used for measuring CH₄ emission, and the other pots for measuring 108 rice productivity traits. Three rice seedlings (28 days old) were transplanted into each pot. 109 Nitrogen was applied as urea, P as calcium superphosphate, and K as potassium chloride in each pot as basal dressing at 165, 88 and 110 kg ha⁻¹, respectively. Side-dressing N fertilizer 110 111 was added at a rate equivalent at 99 kg ha⁻¹ at the tillering stage. During the rice growth 112 period, 2–3 cm water layer overlying the soil surface was maintained.

113

114 *Experiment* 2

115 In this field experiment, we determined the effect of a high-yielding cultivar on CH₄ emissions 116 in a realistic setting. We planted rice seedlings in two adjacent fields at the Jiangpu Farm: one previously fallow field (we will refer to this treatment as "fallow" from now on) with low C 117 118 content, and one paddy field with high C content. Soil properties are reported in Table 1. We 119 used two cultivars that were both commonly grown at the experimental site and differed 120 strongly in biomass and yield: the high-yielding Yangdao 6 (HY), and the low-yielding 121 Ninging 1 (LY). The field experiment was conducted in two adjacent fields with six 122 replicates (3 m \times 4 m in plot size) for each of the soil \times cultivar treatment combination. As a 123 basis for comparison, we also included unplanted plots in our experimental design. The paddy 124 field was in a continuous wheat-rice rotation with adequate plant residues and high labile C 125 content, whereas the fallow field had been fallow for six years prior to the experiment. Few 126 weeds grew in the fallow field and were removed before rice planting.

127Rice seedlings were transplanted at a hill spacing of $0.25 \text{ m} \times 0.20 \text{ m}$ on June 30, 2014.128Nitrogen fertilizer (urea) was applied at 225 kg N ha⁻¹, of which 30% was applied before129planting, another 30% at tillering and the remaining 40% at panicle initiation. Phosphorus130fertilizer (calcium superphosphate) and K (potassium chloride) were applied as the basal131fertilizer at the same rate of 65 kg ha⁻¹. A water layer was kept 4-5 cm above the soil surface132during the pre-anthesis period, while alternate wetting and drying irrigation was applied

134

135 Experiment 3A

In this pot experiment, we used the same soils and cultivars as in field experiment 2. Soils
were collected from each field and sieved (6 mm mesh size) to remove stones. Plastic pots
(height, 25 cm; diameter, 24 cm) were filled with 7 kg of soil. A nylon mesh bag (diameter, 8
cm; height, 10 cm; mesh size, 37 µm) was placed in the center of each pot to create two soil

140 compartments, i.e. the central rooted compartment and the outside non-rooted compartment
141 (Ma *et al.*, 2010). Twenty-five pots were prepared for each cultivar in each soil. Five pots
142 were used for CH₄ emission measurements, and the remaining pots were used for measuring
143 plant traits and soil properties. Two healthy rice seedlings were planted in the root bag. Other
144 management practices were similar as described in experiment 1.

145

146 *Experiment 3B*

147 Using the same experimental approach in Experiment 3A, we also measured CH₄ emissions

148 from fallow soil with and without wheat straw incorporation for both the HY and LY

149 cultivars. Wheat straw incorporation is a widely applied management practice in rice

agriculture (Singh et al., 2008) that strongly increases the amount of soil labile C (Liu et al.,

151 2014). Before the experiment began, fresh wheat straw was chopped and ground into 5-10

mm segments that were incorporated into the soil in each pot at a rate equivalent to 6 t ha⁻¹. A

153 water layer of 4–5 cm was kept during the pre-anthesis period, while alternate wetting and

154 drying irrigation was applied during the post-anthesis period.

155

156 Sampling and measurement methods

Methane emissions in all experiments were measured using the static closed chamber method
(Zou *et al.*, 2005) at 7-day intervals. Methane concentrations were measured by a gas
chromatograph (7890A, Agilent Technologies Inc., USA) equipped with a flame ionization
detector.

161 Dissolved organic C (DOC), root biomass and porosity, and soil methanogenic and 162 methanotrophic gene abundances in experiment 3 were measured on the 55th (Part A) and 163 45th day (Part B) after transplanting, when CH₄ emissions were relatively high and 164 significantly different between the two cultivars. Soil pore water was collected from the root 165 bag compartments using Rhizonsamplers (SMS, Eijkelkamp, Netherlands). About 2 mL soil solution was extracted using a 40 mL vacuum vial to flush and purge the sampler before
sampling, and about 20 mL of soil solution was drawn into another vial. All the sampling
vials were equilibrated by filling them with pure N₂ gas and 5 mL gas of the headspace was
analyzed for CH₄ (Krüger *et al.*, 2001). The solutions were passed through 0.45 µm
membrane filter and analyzed for DOC by a TOC analyzer (multi N/C UV, Analytik Jena AG,
Germany).

172 Fresh soil samples were collected from the rooted compartment. Soil DNA was 173 extracted from 0.25 g soil using a Power Soil DNA Isolation Kit (MoBio, USA). The copy 174 numbers of *mcr*A genes, which represent the abundances of methanogenic archaea in soil, 175 were quantified using the primer pair mcrAf/mcrAr (Luton et al., 2002). Two forward primers 176 of MB10 γ and MB9 α and their common reverse primer 533r were used to quantify the 16S 177 rRNA gene copy numbers of the type I and type II methanotrophs, respectively (Henckel et 178 al., 1999). The quantitative real-time PCR was performed using a Mastercycler ep realplex 179 instrument (Eppendorf, Hamburg, Germany). After sampling the soil, roots were washed with 180 tap water. Rice root porosity (% gas volume/root volume) was measured by the pycnometer 181 method (Jensen et al., 1969). Aboveground biomass and grain yield were measured at harvest. 182 Rice plants were oven-dried at 105 °C for 30 min followed by drying at 70 °C to achieve a 183 constant weight.

184

185 Statistical analysis

186 Correlations between rice plant traits (e.g. root biomass, aboveground biomass, and grain

187 yield) and total seasonal CH₄ emission were analyzed in experiment 1. We also analyzed the

188 correlation between the relative aboveground biomass and relative CH₄ emissions. The

189 relative aboveground biomass and CH₄ emissions were calculated (R):

190 R = xt / xc,

191 where *xt and xc* are the values of the variables (biomass and CH₄ emissions) for a cultivar and

- 192 the lowest values in each soil, respectively. Analysis of variance (two way ANOVA) and
- 193 independent-sample t test for a given soil were performed in experiment 2 and 3. All analyses
- 194 were performed with the statistical package SPSS 18.0. Differences between cultivars were
- 195 considered significant at p < 0.05.
- 196
- 197 Meta-analysis

198 We conducted a literature survey of peer-reviewed papers related to rice cultivars and CH₄

- emission. Peer-reviewed papers published both in English and in Chinese before June 2016
- 200 were collected from the Web of Science and the China National Knowledge Infrastructure.
- 201 We collated studies that met the following criteria:
- 202 (1) soil organic C, rice biomass at harvest and CH₄ emissions were reported simultaneously,
- 203 (2) CH₄ fluxes had to be measured for an entire rice growth period,
- (3) if only two cultivars were used in an experiment, the differences in biomass between thetwo cultivars had to be at least 5%, and
- 206 (4) rice was grown in paddy soils (i.e. studies on fallow soils were excluded).

207 In total, we found 18 published papers including 93 observations from 21 sites 208 (Table 2, Data S1 and S2). For each experiment in our dataset, the rice cultivar with the 209 lowest biomass was taken as the control. We tabulate yield data if they were available, but this 210 was not a prerequisite for inclusion of the experiment in our dataset. We included separate 211 observations of rice cultivar effects from a single study site under different experimental 212 treatments (that is, in multi-factorial studies). Observations from different years within the 213 same experiment were also included as separate observations. For each experiment in the 214 dataset, the rice cultivar with the lowest biomass was taken as the control treatment. We 215 quantified the effects of cultivars with high biomass by calculating the natural logarithm of 216 the response ratio (R): $\ln(R) = \ln (xt / xc)$, where xt and xc are the values of the variables 217 (biomass, yield, HI, or CH₄ emissions) for a cultivar with high biomass and for the control

cultivar, respectively (Hedges *et al.*, 1999). In addition to ln(R), we also used the absolute change in CH₄ emission (Δ CH₄) as effect size to assess the effect of rice cultivars with high biomass on CH₄ emission:

 $221 \qquad \Delta CH_4 = T - C,$

where T and C are the cumulative CH₄ emissions during the growing season of rice cultivars with high biomass and of the control, respectively. Only data collected from field experiments was used in this Δ CH₄ analysis.

225 Three outliers of ln (R_{CH4}) with the largest absolute values (-1.02, -1.03, and 0.90) 226 were identified by the descriptive statistics-explore of the statistical package SPSS 18.0. 227 These three observations and the corresponding observations for ln (R_{biomass}), ln (R_{vield}), ln 228 (R_{HI}), and Δ CH₄ were excluded from further analysis. Because some studies did not report 229 yield, the number of observations for biomass, yield, and CH₄ was not equal. In general, meta-230 analyses assume data independence. This assumption was violated by including more than one 231 observation from a single study, when multiple cultivars within the same study shared the 232 same control treatment. To examine the influence of non-independence, we averaged the 233 effect sizes of different cultivars from the same study in order to make sure that only one 234 comparison was used (Data S2).

We used MetaWin 2.1 (Rosenberg *et al.*, 2000) to generate mean effect sizes and 95% bootstrapped confidence intervals (95% CIs) (4,999 iterations). Because standard deviation values were not reported for most of the observations, we performed our analysis on unweighted effect sizes and on effect sizes that were weighted by replication (Hungate *et al.*, 2009). To compare the differences in effect sizes among soil organic C, soil organic C content was divided into three categories: ≤ 8 g kg⁻¹, 8-12 g kg⁻¹, and >12.0 g kg⁻¹. The mean effect sizes for experimental classes were considered to be significantly

different from each other if their 95% CIs did not overlap, and were significantly different
 from zero if the 95% CI did not overlap with zero. We used randomization tests included in

MetaWin to test for significant differences between study categories. To ease interpretation, the results of this meta-analysis on ln(R) were back-transformed and reported as percentage changes ((R-1) ×100). Results using the different weighting functions were qualitatively similar (Table S1). In the main report, we provide results of the analyses on effect sizes that were weighted by replication.

249

250 *Extrapolation*

251 To scale up our results, we first determined which experimental conditions were most 252 representative of realistic rice paddy systems. Since most rice cropping systems and paddy 253 soil types in the world can be found in China, we took China as a case for assessing the 254 impact of rice cultivar on paddy CH₄ emissions. Based on China's second state soil survey 255 completed in the early 1980s, the arithmetic mean organic C content in the top 15 cm paddy soils was 16.5 g kg⁻¹ (See Fig. S1), and the mean weighted by area was 14.2 g kg⁻¹ (Xie *et al.*, 256 2007). The percentages of the observations with $\leq 8 \text{ g kg}^{-1}$, 8-12 g kg⁻¹, and >12.0 g kg⁻¹ 257 258 organic C content to the total observations are 2.7%, 16.1%, and 81.2%, respectively. 259 Based on the data from the second state soil survey in China (Fig. S1) and the meta-260 analysis, we estimated the effect of increasing biomass on paddy CH₄ emission in China (E): 261 $E = \Sigma (ECi / EBi \times Wi),$ 262 where ECi is the mean effect size of high biomass rice cultivars on CH₄ emissions (%) in 263 the *i*th soil in the meta-analysis, *EBi* is the mean effect size of high biomass rice cultivars on 264 biomass (%) in the *i*th soil in the meta-analysis, and *Wi* is the fraction of the area for the *i*th 265 soil to total paddy soil area. We estimated Wi as the ratio of the number of observations in *i*th 266 soil from the soil survey to the number of total observations from the soil survey.

267

268 **Results**

269 *CH*₄ fluxes

As expected, we found in experiment 1 that CH₄ emissions were higher in the soil with high
labile C content (Fig. 1). In the soil low in labile C, plant productivity was positively
correlated with seasonal cumulative CH₄ emissions (Fig. 1a). However, we found the opposite
relationship for the soil high in labile C (Fig. 1b): as the productivity and yield of the cultivar
increased, cumulative CH₄ emissions declined. For every 10% increase in rice aboveground
biomass, CH₄ emissions declined by 10.3% (Fig. S2).

In experiment 2, we found that in the fallow field, CH₄ emissions were 0.5 mg m⁻² h⁻¹ higher for the high-yielding cultivar compared to the low-yielding cultivar (Fig. 2a; Fig. S3). However, in the paddy field, the opposite pattern occurred, and the high-yielding cultivar reduced CH₄ emissions by 4.6 mg m⁻² h⁻¹ compared to the low-yielding cultivar (Fig. 2a; Fig. S3).

281 Experiment 3A confirmed that the cultivars similarly affected CH₄ emissions in the 282 microcosms as they did under field conditions, with HY increasing CH₄ emissions in the 283 fallow soil, but reducing them in the paddy soil (Fig. 2b, Fig. S4). The results of experiment 284 3B indicates that in the fallow soil without wheat straw incorporation, CH₄ emissions for the high-yielding cultivar were 1.0 mg m⁻² h⁻¹ higher compared to the low-yielding cultivar (Fig. 285 286 2c; Fig. S5). With straw incorporation however, the pattern reversed: CH₄ emissions for the high-yielding cultivar were 9.2 mg m⁻² h^{-1} lower than for the low-yielding cultivar (Fig. 2c; 287 288 Fig. S5). These results strongly suggest that the difference in the effect of high-yielding 289 cultivars on CH₄ emissions between fallow and paddy soils in experiments 2 and 3A are due 290 to differences in labile soil C. Taken together, these three experiments provide conclusive 291 evidence that high-yielding cultivars slightly increase CH₄ emissions in low C soils, but 292 greatly reduce CH₄ emissions in high C soils.

293

294 Soil properties and plant traits

In experiment 3A, we found that the high-yielding cultivar only stimulated methanogens in the fallow soil (Fig. 3a), not in the paddy soil. Similarly, in experiment 3B the high-yielding cultivar only stimulated methanogens in the fallow soil without straw incorporation (Fig. 3b). By contrast, soil methanotrophs were significantly more abundant in the presence of the highyielding cultivar than for the low-yielding rice cultivar in the paddy soil and in the fallow soil with straw incorporation (Fig. 3c and Fig. 3d). In other words, in the high C soils, the highyielding rice cultivar enhanced the abundance of microorganisms that consume CH₄.

302 In experiment 3A, root biomass and DOC values were significantly higher for HY 303 than for LY in both the fallow soil and the paddy soil (Table 3). Similar results were found in 304 the experiment 3B, although straw addition reduced root biomass for both rice cultivars. 305 These results suggest that the high-yielding cultivar enhanced C input to all soils. The root 306 porosity of HY was significantly higher compared to LY in both soils in experiment 3A 307 (Table 3). Similarly, in experiment 3B root porosity was significantly higher for HY than for 308 LY in both the fallow soil and the fallow soil with straw. Straw addition significantly reduced 309 root porosity (P < 0.01). The yield of Yangdao 6 was between 37.2 and 91.8 % higher than for 310 Ninjing 1 in experiment 3. In comparison, the yield of the highest yielding cultivar in 311 experiment 1 was 62.7% and 51.9% higher than that of Ninjing 1 for the low and high soil C 312 soil, respectively (Fig.1). Thus, even though Yangdao 6 was not included in experiment 1, its 313 yield increase relative to Ninjing 1 is comparable to that of other high-yielding cultivars 314 included in our study.

315

316 *Meta-analysis and extrapolation*

317 Our meta-analysis confirmed that on average, rice cultivars with high biomass significantly

- 318 increased CH₄ emissions from lower organic C soils ($\leq 8 \text{ g kg}^{-1}$), but significantly reduced
- 319 CH₄ emissions from higher organic C soils (>12 g kg⁻¹) (Fig. 4). High biomass rice cultivars
- 320 increased yields in all soil organic C classes to a similar extent (Table S1). The average

increase in biomass for studies included in our meta-analysis were 29.8% and 25.6% in the
low C soil and high C soil, respectively (table S1). The meta-analysis of independent data
showed the same trends as the analysis on non-independent data (Table S1), suggesting the
robustness of our results.

Based on the soil survey and our meta-analysis, we estimated the effect of highyielding cultivar breeding strategy on Chinese paddy CH₄ emission by calculating an area weighted effect size. Accounting for the percentage of Chinese rice paddies with ≤ 8 g kg⁻¹, 8-12 g kg⁻¹ and >12.0 g kg⁻¹ organic C contents, we estimated the effect per unit biomass enhancement on CH₄ emissions to be -0.71. In other words, increasing plant biomass by 10% can reduce annual CH₄ emission from Chinese rice agriculture by 7.1%.

331

332 Discussion

333 All our experiments and our meta-analysis show that high-yielding cultivars slightly increase 334 CH₄ emissions in low C soils, but greatly reduce CH₄ emissions in high C soils. Why does the 335 effect of high-yielding cultivars on CH₄ emissions depend on soil C availability? The 336 production of CH₄ is primarily determined by substrate availability (Conrad, 2007), which 337 was enhanced by the high-yielding cultivar, as indicated by both higher root biomass and 338 higher dissolved organic C content in soil pore water of the high-yielding rice cultivar in all 339 the soils used in our experiments. This reflects the higher root productivity of the high-340 yielding cultivar, providing increased substrate availability for CH₄ production through root 341 exudates (Huang *et al.*, 1997). Still, net CH_4 emissions from rice paddies are determined by 342 the balance between the activities of methanogenic archaea, the microorganisms that produce 343 CH₄, and methanotrophic bacteria, the microorganisms that consume CH₄ (Conrad, 2007). 344 Changes in the activities of either microbial group could explain the decline in net emissions 345 observed with the higher-yielding cultivars on high C soils.

346 Methane oxidation and methanotrophic growth are controlled by CH₄ and O₂ 347 availability (Hanson and Hanson, 1996; Conrad, 2007). We propose that the higher root 348 biomass and root porosity of the high-yielding cultivar (Table 3) facilitated O₂ transport into 349 the rhizosphere, stimulating CH₄ oxidation (Ma et al., 2010). This mechanism is particularly 350 important in high C soils, where O₂ is more likely to be limiting (Ma et al., 2010). By 351 contrast, in the fallow soil without straw incorporation, methanotrophic growth was likely 352 limited by low CH₄ availability, especially for the Type II methanotrophs (Hanson and 353 Hanson, 1996; Conrad, 2007). Indeed, CH₄ oxidation in paddy soils only occurs at CH₄ 354 concentrations \geq 500 p.p.m.v. (Cai *et al.*, 2016), far higher than what was found in the fallow 355 soil in experiment 3B. Thus, our results suggest that rice cultivars affect net CH₄ emissions by 356 altering the availabilities of resources that affect microorganisms that both produce and 357 consume CH₄, and that the soil context determines the direction of the effect: high-yielding 358 rice cultivars promote CH₄ production and emissions by increasing C substrate availability for 359 methanogens when soil C content is low, but facilitate CH₄ oxidation by increasing O₂ 360 transport and promoting methanotrophic organisms when soil C availability is high. 361 The generality of our findings were further confirmed by the results of our meta-362 analysis. We can only speculate about the mechanisms underlying the mitigation effect of 363 high-yielding cultivars on CH₄ emissions in our meta-analysis. Indeed, high yielding rice 364 cultivars differ from low yielding cultivars in many different ways that could potentially affect 365 CH₄ emissions. For instance, compared to low yielding cultivars, high yielding cultivars have 366 been shown to increase allocation to panicles (Richards, 2000; Jiang et al., 2016), and to differ 367 in plant growth parameters (Gogoi et al., 2008), root exudation (Lu et al., 2000), and root 368 oxidation activities (Zhang et al., 2009). However, our own data show that high-yielding 369 cultivars increased root porosity, root biomass and methanotrophic activity across multiple 370 independent experiments. These data suggest that the effect of high-yielding cultivars on 371 O₂ transport may be general, occurring under a wide range of environmental

372 conditions and explaining the pattern found across the experiments synthesized in the meta-373 analysis.

374 In our extrapolation, we estimated the effects of a further 10% increase in plant 375 biomass. This represents a realistic scenario: plant breeding efforts have increased the biomass 376 of super rice cultivars in China by about 25% from 2000 to 2015 and are expected to increase 377 a further 10% by 2020 (Peng et al., 2008; Yuan, 2015). In absolute terms, the reduction in the 378 CH₄ emissions caused by rice cultivars with high yield in high organic C soils was an order of 379 magnitude larger than the emission increment in the low organic C soils (Table S1). 380 Moreover, organic C of China's paddy soils has increased by 7.5% from 1979-1982 to 2007-381 2008 (Yan et al., 2011) and will likely continue to increase due to the increasingly common 382 management practice of crop straw incorporation (Singh et al., 2008; Liu et al., 2014). Thus, 383 our estimate of a 7.1% reduction in CH₄ emissions due to high-yielding cultivars is 384 conservative, and the real effect may be larger. 385 Our findings suggest that by switching to high-yielding cultivars, CH₄ emissions from rice 386 agriculture can be reduced substantially. Greenhouse gas emissions from rice 387 paddies will likely be exacerbated because of rising levels of atmospheric CO₂ and climate 388 change (van Groenigen et al., 2011; van Groenigen et al., 2013), further underlining the 389 importance of mitigation measures. However, it is still unclear whether rice cultivar 390 improvement interacts with other agronomic practices (e.g. irrigation, tillage and fertilizer 391 management) to influence CH₄ emissions. These interactions represent a knowledge gap that 392 needs to be addressed to determine the effectiveness of adopting high-yielding cultivars to 393 mitigate CH₄ emissions.

Two limitations of our study must be noted. First, our experiments lasted for one growing season. However, some of the effects of high-yielding cultivars on soil C input will only become apparent in long-term experiments, when biomass produced in one season contributes to soil C input in the next season. Indeed, numerous studies (e.g. Feng et al., 2013;

our study) show that rice straw incorporation strongly stimulate CH₄ emissions; increased
biomass and straw production with high yielding cultivars would enhance these effects. On
the other hand, increased straw input will increase soil C availability, and the mitigation effect
of high-yielding cultivars on CH₄ emissions become more pronounced in high C soils.
Clearly, long-term studies are needed to confirm whether the mitigation effects of highyielding cultivars persist over time.

Second, the microbiological analyses in our study are all based on single
measurements in time. Microbial communities are dynamic, so the microbiological data
presented here should be viewed accordingly. The data support our hypothesis that highyielding cultivars reduce CH₄ emissions by stimulating oxygen transport into the soil, but
future studies should include a time series component to confirm whether effects of highyielding rice cultivars on methanotrophs and methanogens persist throughout the growing
season.

411 Maintaining food security in the face of population growth and climate change is one 412 of great challenges facing mankind today (Alexandratos and Bruinsma, 2012). Food security 413 can be enhanced through agricultural intensification, but measures that increase crop yields 414 often increase greenhouse gas emissions too (Tilman, 1999). Here, we show that agricultural 415 intensification can go hand in hand with greenhouse gas mitigation. Other mitigation practices 416 advocated to curb CH₄ emissions from rice paddies include mid-season drainage, intermittent 417 irrigation, no-till, and the use of alternative fertilizers (Hussain et al., 2015; Linquist et al., 418 2015; Zhao et al., 2016). However, these practices can result in yield losses (Pittelkow et al., 419 2015), are labor-intensive, and their applicability varies among rice cropping systems and 420 countries (Bodelier, 2015). In contrast, rice cultivar improvement may be a win-win strategy, 421 as it simultaneously decreases CH₄ emissions and increases grain yield. Although seeds of 422 higher yielding cultivars will be more expensive, farmers benefit where an increase in grain 423 yield exceeds extra cost and society benefits through the reduction in greenhouse gases.

424	Considering the dominance of small households in most rice production areas (Zhang et al.,
425	2016), the use of high-yielding cultivars may therefore be accepted sooner and implemented
426	more efficiently than other mitigation practices. Along with other mitigation efforts, future
427	policy measures aimed at reducing CH4 emissions from rice cultivation should consider the
428	use of high-yielding cultivars.
429	
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595	Supporting Information
596 597	Additional Supporting information may be found in the online version of this article:
598 599	Data S1. This data set lists all experimental non- independent observations from rice cultivar experiments that were used for the meta-analysis on CH ₄ fluxes, yield, and harvest index. It
600 601 602	how the data were extracted from each publication.
603	Data S2. This data set lists all experimental independent observations from rice cultivar
604	experiments that were used for the meta-analysis on CH4 fluxes, yield, and harvest index. It
605	briefly summarizes the experimental conditions under which the observations were made, and
606	how the data were extracted from each publication.
607	
608	Table S1. Results of a meta-analysis on the effects of high biomass cultivars on biomass,
609	yield, harvest index, and CH ₄ emission. Results are shown for categories based on the
610	following organic soil C contents: $\leq 8 \text{ g kg}^{-1}$, 8-12 g kg ⁻¹ , and > 12 g kg ⁻¹ .
611	
612	Fig. S1. Soil organic C content in the top layer of paddy soils in China.
613	
614	Fig. S2. The relation between relative aboveground biomass and relative seasonal cumulative
615	CH ₄ emissions in a soil with low labile C content (a) and a soil with high labile C content (b)
616	across 33 rice cultivars.
617	
618	Fig. S3. CH ₄ emissions from unplanted soils, soils planted with a high-yielding rice cultivar
619	and soils planted with low-yielding rice cultivar under field conditions.
620	
621	Fig. S4. CH ₄ emissions from unplanted soil, soils planted with a high-yielding rice cultivar
622	and soils planted with low-yielding rice cultivar under pot conditions.
623	
624	Fig. S5. CH ₄ emissions from unplanted microcosms, microcosms planted with a high-yielding
625	rice cultivar, and microcosms planted with a high-yielding rice cultivar.
626	
627	

628 Table 1 Main properties of the tested soils and rice cultivars used in our st
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	Experiment 1		Experiments 2 and 3	
	Paddy soil	Dried soil	Paddy soil	Fallow soil
Soil organic C (g kg ⁻¹)	17.8	17.5	23.2	12
Soil labile C (g kg ⁻¹)	3.6	1.4	6.5	1.0
Total N (g kg ⁻¹)	2.1	2.0	2.3	1.5
Total P (g kg ⁻¹)	0.6	0.6	0.9	0.7
Total K (g kg ⁻¹)	13.8	14.0	8.2	11
Alkaline hydrolysis N (mg kg ⁻¹)	96.8	94.1	85.5	86.1
Available P (mg kg ⁻¹)	28.1	16.6	20.9	22.5
Available K (mg kg ⁻¹)	244.5	165.0	206.1	138.3
Soil pH	6.8	6.7	6.5	6.9
Rice cultivars	Eryou 084, Fengliangyouxiang 1, Fengyuan 299, Guizhannong, Guodao 1, Hezhanmei, Huaidao 9, Huailiangyou 527, Huiliangyou 6, Jijing 88, Liaoxing 1, Liaoyou 1052, Liaoyou 5218, Longdao 5, Nei2you 6, Ningjing 1, Ningjing 3, Peizafengtai, Qianchonglang 2, Shengnong 016, Shengnong 265, Shengnong 9816, Wuyou 308, Wuyunjing 24, Yangjing 4038, Xindao 18, Xinliangyou 6, Xinliangyou 638, Yliangyou 1, Yliangyou 302, Yangliangyou 7, Yuxiangyouzhan,			

Table 2 Overview of the rice cultivar studies included in our meta-anal	lysis.	
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Site	Country	SOC (g kg ⁻¹)	# rice cultivars	n	Experimental condition	Mean CH ₄ effect (%)	Mean yield effect (%)	Reference
Aichi	Japan	9.8	3	2	Pot	12.1	NA	Watanabe et al., 2001
Assam	India	8.0	2	10	Field	7.8	NA	Gogoi et al., 2005
Assam	India	6.0	2	10	Field	16.1	NA	Gogoi et al., 2005
Beijing	China	9.9	4	4	Field	19.8	58.5	Wang et al., 2000
Beijing	China	10.0	3	4	Field	123.9	-11.0	Xu et al., 1999
Cuttack	India	6.6	2	3	Field	-1.8	32.6	Datta et al., 2009
Cuttack	India	7.6	10	3	Field	-2.0	66.4	Satpathy et al., 1998
Danyang	China	19.6	10	3	Field	-22.0	8.5	Zhang <i>et al.</i> , 2015
Hangzhou	China	22.4	6	3	Field	-10.4	4.3	Lu et al., 2000
Java	Indonesia	4.8	2	3	Field	11.6	5.6	Setyanto et al., 2000
Jiangdu	China	15.0	2	4	Field	-11.8	-3.3	Tang <i>et al.</i> , 2015
Jinxian	China	25.0	9	3	Field	-10.3	11.4	Zhang <i>et al.</i> , 2015
Laguna	Philippine	12.0	5	4	Field	-25.0	NA	Wassmann et al., 2000
Nanjing	China	17.8	2	3	Pot	-26.0	35.0	Wang et al., 2013
New Delhi	India	4.5	6	3	Field	42.1	6.3	Mitra <i>et al.</i> , 1999
New Delhi	Indian	4.1	3	NA	Field	6.0	14.5	Jain <i>et al.</i> , 2000
Sacheon	Korea	9.8	8	3	Field	-5.8	-1.2	Gutierrez et al., 2013
Shenyang	China	33.7	12	3	Field	-32.7	25.4	Zhang <i>et al.</i> , 2015
Tokyo	Japan	36.3	2	3	Field	-37.3	47.6	Win et al., 2016
Tsukuba	Japan	7.5	4	3	GC	36.6	35.8	Lou et al., 2008
Varanasi	India	7.2	3	3	Field	-10.5	5.8	Singh et al., 1999

633 NA, not reported; GC, growth chamber

Table 3. Plant traits, dissolved organic C and CH₄ in soil pore water, and Type I and II

methanotrophs for a high-yielding (HY) and a low-yielding (LY) rice cultivar in experiment

Experiment 3A	Fallow soil		Paddy s	soil
-	HY	LY	HY	LY
Root biomass (g plant ⁻¹)	$5.6 \pm 0.2^{**}$	3.6 ± 0.1	$7.1 \pm 0.9*$	4.2 ± 0.3
Root porosity (%)	$39.6 \pm 2.7*$	33.3 ± 1.1	$43.4 \pm 0.7*$	37.7 ± 1.5
Dissolved organic C (mg L ⁻¹)	$82.0 \pm 4.2^{**}$	68.5 ± 1.7	$132.3 \pm 3.7*$	111.6 ± 3.4
Type I methanotrophs copies $(10^7 \text{ copies } g^{-1} \text{ dry soil})$	4.7 ± 0.4	5.1 ± 0.6	$7.9 \pm 0.3*$	6.6 ± 0.2
Type II methanotrophs 10 ⁷ copies g ⁻¹ dry soil)	6.9 ± 0.4	6.8 ± 0.6	$10.2\pm0.4*$	7.0 ± 0.2
Aboveground biomass (g plant ⁻¹)	$40.2 \pm 1.6^{**}$	26.4 ± 1.6	$55.8 \pm 2.0 **$	41.5 ± 1.8
Grain yield (g plant ⁻¹)	$21.1 \pm 1.0^{**}$	11.0 ± 0.8	$27.2 \pm 1.8 **$	18.9 ± 0.7

Experiment 3B	Fallow soil with	nout straw	Fallow soil with straw		
-	HY	LY	HY	LY	
Root biomass (g plant ⁻¹)	$4.0 \pm 0.4*$	2.5 ± 0.3	$3.3 \pm 0.2^{**}$	2.0 ± 0.1	
Root porosity (%)	$36.7 \pm 2.8*$	28.5 ± 0.7	$23.1 \pm 0.1*$	17.9 ± 0.5	
Dissolved organic C (mg L ⁻¹)	$74.5 \pm 1.5*$	63.9 ± 0.6	$123.1 \pm 2.4 **$	101.0 ± 3.1	
CH ₄ in soil pore water (p.p.m.v)	$193.3 \pm 15.4*$	87.6 ± 10.5	$1553.9 \pm 31.6^*$	3397.2 ± 521.5	
Type I methanotrophs copies	4.1 ± 0.1	4.2 ± 0.2	$6.2 \pm 0.8*$	4.7 ± 0.2	
$(10^7 \text{ copies } \text{g}^{-1} \text{ dry soil})$					
Type II methanotrophs	6.8 ± 0.6	6.7 ± 0.4	$11.45 \pm 0.9*$	8.3 ± 0.4	
$(10^7 \text{ copies g}^{-1} \text{ dry soil})$					
Aboveground biomass (g plant ⁻¹)	$38.6 \pm 0.7 **$	28.1 ± 0.7	$35.1 \pm 1.5 **$	25.1 ± 0.3	
Grain yield (g plant ⁻¹)	$19.9 \pm 0.4 **$	14.5 ± 0.8	$17.9 \pm 0.7 **$	13.0 ± 0.4	

Mean \pm standard error (n=5). * and ** indicate significant differences between cultivars at p < 0.05 and 0.01,

639 640 respectively.





Fig. 1 Relationships between plant productivity traits (i.e. root biomass, aboveground

661	biomass, and yield) and seasonal cumulative CH ₄ emissions across 33 rice cultivars. (a) soil
662	CH ₄ emissions vs. plant productivity in a soil with low labile C content. Cumulative CH ₄
663	emissions were positively correlated with root biomass ($r^2 = 0.34$), grain yield ($r^2 = 0.29$) and
664	above ground biomass ($r^2 = 0.38$); (b) soil CH ₄ emissions vs. plant productivity in a soil with
665	high labile C content. Cumulative CH4 emissions were negatively correlated with root
666	biomass ($r^2 = 0.30$), grain yield ($r^2 = 0.39$) and above ground biomass ($r^2 = 0.46$). All
667	correlations were significant at $p < 0.01$. The results for cultivar Ninjing 1 (LY in experiments

668 2 and 3) are indicated by black symbols.





Fig. 2 CH₄ emissions from a high-yielding (HY) and a low-yielding (LY) rice cultivar, as affected by soil C contents. (a) CH₄ emissions from a fallow soil (low soil C content) and a paddy soil (high soil C content) under field conditions. (b) CH₄ emissions from a fallow soil and a paddy soil under pot conditions. (c) CH₄ emissions from fallow soil with and without straw incorporation under pot conditions. Error bars represent standard error (n = 6 for the field experiment, n = 5 for the pot experiment). ** indicates significant difference between cultivars at p < 0.01.





678 Fig. 3 Quantification of methanogens and methanotrophs under a high-yielding (HY) and a 679 low-yielding (LY) rice cultivar, as affected by soil C contents. (a) Quantification of 680 methanogens in a fallow soil (low soil C content) and a paddy soil (high soil C content). (b) 681 Quantification of methanogens in a fallow soil with and without straw incorporation. (c) 682 Quantification of methanotrophs in a fallow soil and a paddy soil. (d) Quantification of 683 methanotrophs in a fallow soil with and without straw incorporation. Error bars represent standard errors (n=5). * and ** indicate significant differences between cultivars at p < 0.05684 685 and 0.01, respectively.



687 Fig. 4 Results of a meta-analysis on the effect of high-biomass rice cultivars on CH₄

688 emissions under different soil organic C contents. Results are based on 33, 25, and 35

 $\label{eq:constraint} 689 \qquad \text{observations for the soil organic C} \leq 8 \text{ g kg}^{-1}\text{, 8-12 g kg}^{-1}\text{, and } > 12.0 \text{ g kg}^{-1}\text{ class, respectively.}$

690 Error bars indicate 95% confidence intervals. The effect of high-yielding cultivars on CH₄

691 emissions differed significantly between experimental classes (p = 0.0002).