Higher yields and lower methane emissions with new rice cultivars

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Abstract
Breeding high-yielding rice cultivars through increasing biomass is a key strategy to meet 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 powerful greenhouse gas. Here, we show in a series of experiments that high-yielding rice cultivars actually reduce CH4 emissions from typical paddy soils. Averaged across 33 rice cultivars, a biomass increase of 10% resulted in a 10.3% decrease in CH4 emissions in a soil with a high carbon (C) content. Compared to a low-yielding cultivar, a high-yielding cultivar significantly increased root porosity and the abundance of methane-consuming microorganisms, suggesting that the larger and more porous root systems of high-yielding cultivars facilitated CH4 oxidation by promoting O2 transport to soils. Our results were further supported by a meta-analysis, showing that high-yielding rice cultivars strongly decrease CH4 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%. Our findings suggest that modern rice breeding strategies for
1 | 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 & 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, van Groenigen, Adviento Borbe, Pittelkow, & van Kessel, 2012). Therefore, sustainable intensification of rice cropping systems requires increasing yields while reducing CH4 emissions (Chen et al., 2014; van Groenigen, van Kessel, & Hungate, 2013; Linquist et al., 2012).

Global rice production can be increased through improving yield potential of rice cultivars; the introduction of high-yielding rice cultivars accounts for almost 50% of the recent yield growth in developing countries (Evensen & Gollin, 2003; Peng, Khush, Virk, Tang, & Zou, 2008). 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 CH4 emissions, as an increase in harvest index with constant plant biomass can decrease the production of root exudates that fuel CH4 production (Su et al., 2015; Van Der Gon et al., 2002). However, the current harvest index of high-yielding rice cultivars is about 0.55, approaching the theoretical upper limit of 0.65 (Hay, 1995; Peng et al., 2008). Therefore, more recent breeding strategies for increasing yields focus on enhancing biomass while maintaining the current harvest index (Cheng et al., 2007; Peng et al., 2008; Richards, 2000; Yuan, 2015).

These latter breeding strategies could stimulate CH4 emissions, because recent photosynthate of rice plants can be a major substrate for CH4 production: with higher biomass production, more substrate could fuel higher CH4 emission rates (Huang, Sass, & Fisher, 1997; Watanabe, Takeda, & Kimura, 1999). The microorganisms that produce CH4, methanogenic archaea, also use substrates that are derived from native soil organic carbon (C) (Conrad, 2007; Watanabe et al., 1999), suggesting that the effect of rice cultivars on CH4 emissions depends on soil C content. Thus, to study the effects of high-yielding cultivars on CH4 emissions and their possible interaction with soil C availability, we conducted three independent but complementary experiments. (i) We used 33 rice cultivars to quantify the relationship between plant production and CH4 emission in two otherwise similar paddy soils with different labile soil C contents, (ii) we determined the effect of a high-yielding cultivar on CH4 emissions in a realistic field setting, and (iii) 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. Finally, to test the generality of our findings, we conducted a meta-analysis of studies that quantified the effect of high-yielding rice cultivars on CH4 emissions.

2 | MATERIALS AND METHODS

2.1 | Experiment 1

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 experiment was conducted under open field conditions at Pailou experimental station, Nanjing Agricultural University, Nanjing City (118.8°E, 32.1°N), China. Thirty-three rice cultivars approved and released since 2001 in China (Table 1) were tested. Both soils in this experiment were collected from the plow layer of the paddy fields at Jiangpu Farm of Nanjing Agricultural University, Nanjing City. The low C soil was stored outdoors for 3 years before being used; soil labile C content was low because most of the plant residues and other labile C in the soil had been oxidized or mineralized. The soil with a high soil labile C content was collected 5 days before the experiment. Soil labile C content was measured by the KMnO4 oxidation method (Blair, Lefroy, & Lisle, 1995). Soil properties are reported in Table 1.

Plastic pots (height, 25 cm; diameter, 24 cm) were filled with 7.0 kg of soil that was sieved (6 mm mesh size) to remove stones. Fifteen pots were prepared for each cultivar with each soil. Three pots were used for measuring CH4 emission, and the other pots for measuring rice productivity traits. Three rice seedlings (28 days old) were transplanted into each pot. Nitrogen was applied as urea, P as calcium superphosphate, and K as potassium chloride in each pot at basal dressing at 165, 88, and 110 kg ha⁻¹, respectively. Side-dressing N fertilizer was added at a rate equivalent at 99 kg ha⁻¹ at the tillering stage. During the rice growth period, 2–3 cm water layer overlying the soil surface was maintained.

2.2 | Experiment 2

In this field experiment, we determined the effect of a high-yielding cultivar on CH4 emissions 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 content, and one paddy field with high C content. Soil properties are reported in Table 1. We used two cultivars that were
both commonly grown at the experimental site and differed strongly in biomass and yield: the high-yielding Yangdao 6 (HY) and the low-yielding Ningjing 1 (LY). The field experiment was conducted in two adjacent fields with six replicates (3 m × 4 m in plot size) for each of the soil × cultivar treatment combination. As a basis for comparison, we also included unplanted plots in our experimental design. The paddy field was in a continuous wheat–rice rotation with adequate plant residues and high labile C content, whereas the fallow field had been fallow for 6 years prior to the experiment. Few weeds grew in the fallow field and were removed before rice planting.

Rice seedlings were transplanted at a hill spacing of 0.25 m × 0.20 m on 30 June 2014. Nitrogen fertilizer (urea) was applied at 225 kg N ha⁻¹, of which 30% was applied before planting, another 30% at tillering, and the remaining 40% at panicle initiation. Phosphorus fertilizer (calcium superphosphate) and K (potassium chloride) were applied as the basal fertilizer at the same rate of 65 kg ha⁻¹. A water layer was kept 4–5 cm above the soil surface during the pre-anthesis period, while alternate wetting and drying irrigation was applied during the postanthesis period.

### Tables

#### Table 1: Main properties of the tested soils and rice cultivars used in our study

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<thead>
<tr>
<th></th>
<th>Experiment 1</th>
<th></th>
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<tbody>
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<td>Paddy soil</td>
<td>Dried soil</td>
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<td>17.5</td>
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<td>1.4</td>
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<tr>
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<td></td>
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<td>0.6</td>
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<tr>
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<td>86.1</td>
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<td>16.6</td>
<td>20.9</td>
<td>22.5</td>
<td></td>
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<tr>
<td>Available K (mg kg⁻¹⁻¹)</td>
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<td>165.0</td>
<td>206.1</td>
<td>138.3</td>
<td></td>
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<td>6.7</td>
<td>6.5</td>
<td>6.9</td>
<td></td>
</tr>
<tr>
<td>Rice cultivars</td>
<td>Eryou 084, Fengliangyouxiang 1, Fengyuan 299, Guizhannong, Guodao 1, Hezhanmei, Hualiao 9, Huiliangyou 527, Huiliangyou 6, Jijing 88, Liaoxing 1, Liaoyou 1052, Liaoyou 5218, Longdao 5, Neiyou 6, Ningjing 1, Ningjing 3, Peizafengtai, Qianchonglang 2, Shengnong 016, Shengnong 265, Shengnong 9816, Wuyou 308, Wuyunjing 24, Yangjing 4038, Xindao 18, Xiniangyou 6, Xiniangyou 638, Yiliangyou 1, Yiliangyou 302, Yangliangyou 7, Yuxiangyouzhan, Zhongzeyou 1</td>
<td>Ningjing 1, Yangdao 6</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

### 2.3 | 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 compartments, that is, the central rooted compartment and the outside nonrooted compartment (Ma, Qiu, & Lu, 2010). Twenty-five pots were prepared for each cultivar in each soil. Five pots were used for CH₄ emission measurements, and the remaining pots were used for measuring plant traits and soil properties. Two healthy rice seedlings were planted in the root bag. Other management practices were similar as described in experiment 1.

### 2.4 | Experiment 3B

Using the same experimental approach in Experiment 3A, we also measured CH₄ emissions from fallow soil with and without wheat straw incorporation for both the HY and LY cultivars. Wheat straw incorporation is a widely applied management practice in rice agriculture (Singh, Shan, Johnson-Beebout, Singh, & Buresh, 2008) that strongly increases the amount of soil labile C (Liu, Lu, Cui, Li, & Fang, 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 water layer of 4–5 cm was kept during the pre-anthesis period, while alternate wetting and drying irrigation was applied during the postanthesis period.

### 2.5 | Sampling and measurement methods

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

Dissolved organic C (DOC), root biomass and porosity, and soil methanogenic and methanotrophic gene abundances in experiment 3 were measured on the 55th (Part A) and 45th day (Part B) after transplanting, when CH₄ emissions were relatively high and
significantly different between the two cultivars. Soil pore water was collected from the root bag compartments using Rhizosamplers (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, Frenzel, & Conrad, 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).

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

### 2.6 Statistical analysis

Correlations between rice plant traits (e.g., root biomass, aboveground biomass, and grain yield) and total seasonal CH₄ emission were analyzed in experiment 1. We also analyzed the correlation between the relative aboveground biomass and relative CH₄ emissions. The relative aboveground biomass and CH₄ emissions were calculated (R):

\[
R = \frac{xt}{xc},
\]

where xt and xc are the values of the variables (biomass and CH₄ emissions) for a cultivar and the lowest values in each soil, respectively. Analysis of variance (two-way ANOVA) and independent sample t-test for a given soil were performed in experiments 2 and 3. All analyses were performed with the statistical package SPSS 18.0. Differences between cultivars were considered significant at p < .05.

### 2.7 Meta-analysis

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 were collected from the Web of Science and the China National Knowledge Infrastructure. We collated studies that met the following criteria:

1. soil organic C, rice biomass at harvest, and CH₄ emissions were reported simultaneously,
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 the two cultivars had to be at least 5%, and
4. rice was grown in paddy soils (i.e., studies on fallow soils were excluded).

In total, we found 18 published papers including 93 observations from 21 sites (Table 2, Data S1 and S2). For each experiment in our dataset, the rice cultivar with the lowest biomass was taken as the control. We tabulate yield data if they were available, but this was not a prerequisite for inclusion in the experiment in our dataset. We included separate observations of rice cultivar effects from a single study site under different experimental treatments (i.e., in multifactorial studies). Observations from different years within the same experiment were also included as separate observations. For each experiment in the dataset, the rice cultivar with the lowest biomass was taken as the control treatment. We quantified the effects of cultivars with high biomass by calculating the natural logarithm of the response ratio (R): ln(R) = ln(xt/xc), where xt and xc are the values of the variables (biomass, yield, HI, or CH₄ emissions) for a cultivar with high biomass and for the control cultivar, respectively (Hedges, Gurevitch, & Curtis, 1999). In addition to ln(R), we also used the absolute change in CH₄ emission (ΔCH₄) as effect size to assess the effect of rice cultivars with high biomass on CH₄ emission:

\[
ΔCH₄ = T - C,
\]

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 were used in this ΔCH₄ analysis.

Three outliers of ln (R(CH₄)) with the largest absolute values (–1.02, –1.03, and 0.90) were identified by the descriptive statistics explore of the statistical package SPSS 18.0. These three observations and the corresponding observations for ln (R(biomass)), ln (R(yield)), and ΔCH₄ were excluded from further analysis. Because some studies did not report yield, the number of observations for biomass, yield, and CH₄ was not equal. In general, meta-analyses assume data independence. This assumption was violated by including more than one observation from a single study, when multiple cultivars within the same study shared the same control treatment. To examine the influence of nonindependence, we averaged the effect sizes of different cultivars from the same study in order to make sure that only one comparison was used (Data S2).

We used MetaWin 2.1 (Rosenberg, Adams, & Gurevitch, 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 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 (\(\frac{R}{C_0}\times 100\)). 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.

### 2.8 Extrapolation

To scale up our results, we first determined which experimental conditions were most representative of realistic rice paddy systems. As most rice cropping systems and paddy soil types in the world can be found in China, we took China as a case for assessing the impact of rice cultivar on paddy \(CH_4\) emissions. Based on China’s second state soil survey 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 Figure S1), and the mean weighted by area was 14.2 g kg\(^{-1}\) (Xie et al., 2007). The percentages of the observations with \(<8\) g kg\(^{-1}\), 8–12 g kg\(^{-1}\), and \(>12.0\) g kg\(^{-1}\) organic C content to the total observations are 2.7%, 16.1%, and 81.2%, respectively.

Based on the data from the second state soil survey in China (Figure S1) and the meta-analysis, we estimated the effect of increasing biomass on paddy \(CH_4\) emission in China (\(E\)):

\[
E = \Sigma \left( \frac{EC_i}{EB_i} \times W_i \right),
\]

where \(EC_i\) is the mean effect size of high biomass rice cultivars on \(CH_4\) emissions (%) in the \(i\)th soil in the meta-analysis, \(EB_i\) is the mean effect size of high biomass rice cultivars on biomass (%) in the \(i\)th soil in the meta-analysis, and \(W_i\) is the fraction of the area for the \(i\)th soil to total paddy soil area. We estimated \(W_i\) as the ratio of the number of observations in the \(i\)th soil from the soil survey to the number of total observations from the soil survey.

### 3 RESULTS

#### 3.1 \(CH_4\) fluxes

As expected, we found in experiment 1 that \(CH_4\) emissions were higher in the soil with high labile C content (Figure 1). In the soil low in labile C, plant productivity was positively correlated with seasonal cumulative \(CH_4\) emissions (Figure 1a). However, we found the opposite relationship for the soil high in labile C (Figure 1b): as the productivity and yield of the cultivar increased, cumulative \(CH_4\)

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<table>
<thead>
<tr>
<th>Site</th>
<th>Country</th>
<th>SOC (g kg(^{-1}))</th>
<th>No of rice cultivars</th>
<th>(n)</th>
<th>Experimental condition</th>
<th>Mean (CH_4) effect (%)</th>
<th>Mean yield effect (%)</th>
<th>Reference</th>
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<tbody>
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<td>Aichi</td>
<td>Japan</td>
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<td>3</td>
<td>2</td>
<td>Pot</td>
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<td>2</td>
<td>10</td>
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<td>NA</td>
<td>Gogoi, Baruah, Gogoi, and Gupta (2005)</td>
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<td>3</td>
<td>GC</td>
<td>36.6</td>
<td>35.8</td>
<td>Lou et al. (2008)</td>
</tr>
<tr>
<td>Varanasi</td>
<td>India</td>
<td>7.2</td>
<td>3</td>
<td>3</td>
<td>Field</td>
<td>−10.5</td>
<td>5.8</td>
<td>Singh, Singh, and Kashyap (1999)</td>
</tr>
</tbody>
</table>

NA, not reported; GC, growth chamber.
emissions declined. For every 10% increase in rice aboveground biomass, CH4 emissions declined by 10.3% (Figure S2).

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

Experiment 3A confirmed that the cultivars similarly affected CH4 emissions in the microcosms as they did under field conditions, with HY increasing CH4 emissions in the fallow soil, but reducing them in the paddy soil (Figures 2a and S3). The results of experiment 3B indicate that in the fallow soil without wheat straw incorporation, CH4 emissions for the high-yielding cultivar were 1.0 mg m$^{-2}$ hr$^{-1}$ higher compared to the low-yielding cultivar (Figures 2c and S5). With straw incorporation, however, the pattern reversed: CH4 emissions for the high-yielding cultivar were 9.2 mg m$^{-2}$ hr$^{-1}$ lower than for the low-yielding cultivar (Figures 2c and S5). These results strongly suggest that the difference in the effect of high-yielding cultivars on CH4 emissions between fallow and paddy soils in experiments 2 and 3A are due to differences in labile soil C. Taken together, these three experiments provide conclusive evidence that high-yielding cultivars slightly increase CH4 emissions in low C soils, but greatly reduce CH4 emissions in high C soils.

### 3.2 Soil properties and plant traits

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

In experiment 3A, root biomass and DOC values were significantly higher for HY than for LY in both the fallow soil and the paddy soil (Table 3). Similar results were found in the experiment 3B, although straw addition reduced root biomass for both rice cultivars. These results suggest that the high-yielding cultivar enhanced C input to all soils. The root porosity of HY was significantly higher compared to LY in both soils in experiment 3A (Table 3). Similarly, in experiment 3B, root porosity was significantly higher for HY than for LY in both the fallow soil and the fallow soil with straw. Straw addition significantly reduced root porosity (P < .01). The yield of Yangdao 6 was between 37.2 and 91.8% higher than for Ninjing 1 in experiment 3. In comparison, the yield of the highest yielding cultivar in experiment 1 was 62.7% and 51.9% higher than that of Ninjing 1 for the low and high soil C soil, respectively (Figure 1). Thus, even though Yangdao 6 was not included in experiment 1, its yield increase relative to Ninjing 1 is comparable to that of other high-yielding cultivars included in our study.

**FIGURE 3** Quantification of methanogens and methanotrophs under a high-yielding (HY) and a low-yielding (LY) rice cultivar, as affected by soil C contents. (a) Quantification of methanogens in a fallow soil (low soil C content) and a paddy soil (high soil C content). (b) Quantification of methanogens in a fallow soil with and without straw incorporation. (c) Quantification of methanotrophs in a fallow soil and a paddy soil. (d) Quantification of 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 < .05 and 0.01, respectively.
TABLE 3  Plant traits, dissolved organic C and CH₄ in soil pore water, and Types I and II methanotrophs for a high-yielding (HY) and a low-yielding (LY) rice cultivar in experiment 3

<table>
<thead>
<tr>
<th>Experiment 3A</th>
<th>Fallow soil without straw</th>
<th>Fallow soil with straw</th>
<th>Paddy soil</th>
</tr>
</thead>
<tbody>
<tr>
<td>Root biomass (g plant⁻¹)</td>
<td>HY</td>
<td>5.6 ± 0.2**</td>
<td>7.1 ± 0.9*</td>
</tr>
<tr>
<td>Root porosity (%)</td>
<td>LY</td>
<td>3.6 ± 0.1</td>
<td>43.4 ± 0.7*</td>
</tr>
<tr>
<td>Dissolved organic C (mg L⁻¹)</td>
<td>HY</td>
<td>82.0 ± 4.2**</td>
<td>132.3 ± 3.7*</td>
</tr>
<tr>
<td>Type I methanotrophs copies (10⁷ copies g⁻¹ dry soil)</td>
<td>LY</td>
<td>68.5 ± 1.7</td>
<td>10.2 ± 0.4*</td>
</tr>
<tr>
<td>Aboveground biomass (g plant⁻¹)</td>
<td>HY</td>
<td>4.7 ± 0.4</td>
<td>7.9 ± 0.3*</td>
</tr>
<tr>
<td>Grain yield (g plant⁻¹)</td>
<td>LY</td>
<td>5.1 ± 0.6</td>
<td>6.6 ± 0.2</td>
</tr>
<tr>
<td>Exp 3B</td>
<td>Fallow soil without straw</td>
<td>Fallow soil with straw</td>
<td></td>
</tr>
<tr>
<td>Root biomass (g plant⁻¹)</td>
<td>HY</td>
<td>4.0 ± 0.4*</td>
<td>3.3 ± 0.2**</td>
</tr>
<tr>
<td>Root porosity (%)</td>
<td>LY</td>
<td>2.5 ± 0.3</td>
<td>2.0 ± 0.1</td>
</tr>
<tr>
<td>Dissolved organic C (mg L⁻¹)</td>
<td>HY</td>
<td>74.5 ± 1.5*</td>
<td>23.1 ± 0.1*</td>
</tr>
<tr>
<td>CH₄ in soil pore water (ppm)</td>
<td>LY</td>
<td>63.9 ± 0.6</td>
<td>17.9 ± 0.5</td>
</tr>
<tr>
<td>Type I methanotrophs copies (10⁷ copies g⁻¹ dry soil)</td>
<td>HY</td>
<td>193.3 ± 15.4*</td>
<td>123.1 ± 2.4**</td>
</tr>
<tr>
<td>Type II methanotrophs copies (10⁷ copies g⁻¹ dry soil)</td>
<td>LY</td>
<td>87.6 ± 10.5</td>
<td>101.0 ± 3.1</td>
</tr>
<tr>
<td>Aboveground biomass (g plant⁻¹)</td>
<td>HY</td>
<td>4.1 ± 0.1</td>
<td>6.2 ± 0.8*</td>
</tr>
<tr>
<td>Grain yield (g plant⁻¹)</td>
<td>LY</td>
<td>4.2 ± 0.2</td>
<td>4.7 ± 0.2</td>
</tr>
</tbody>
</table>

Mean ± standard error (n = 5). * and ** indicate significant differences between cultivars at p < .05 and .01, respectively.

3.3  Meta-analysis and extrapolation

Our meta-analysis confirmed that on average, rice cultivars with high biomass significantly increased CH₄ emissions from lower organic C soils (<8 g kg⁻¹), but significantly reduced CH₄ emissions from higher organic C soils (>12 g kg⁻¹) (Figure 4). High biomass rice cultivars 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 was 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 nonindependent data (Table S1), suggesting the robustness of our results.

Based on the soil survey and our meta-analysis, we estimated the effect of high-yielding 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%.

4  DISCUSSION

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

Methane oxidation and methanotrophic growth are controlled by CH₄ and O₂ availability (Conrad, 2007; Hanson & Hanson, 1996). We propose that the higher root biomass and root porosity of the high-yielding cultivar (Table 3) facilitated O₂ transport into the rhizosphere, stimulating CH₄ oxidation (Ma et al., 2010). This mechanism is particularly important in high C soils, where O₂ is more likely to be limiting (Ma et al., 2010). By contrast, in the fallow soil without straw incorporation, methanotrophic growth was likely limited by low CH₄ availability, especially for the Type II methanotrophs (Conrad, 2007; Hanson & Hanson, 1996). Indeed, CH₄ oxidation in paddy soils only occurs at CH₄ concentrations ≥500 p.p.m.v. (Cai, Zheng, Bodenner, Conrad, & Jia, 2016), far higher than what was found in
of high-yielding cultivars on CH4 emissions differed significantly respectively. Error bars indicate 95% confidence intervals. The effect when soil C content is low, but facilitate CH4 oxidation by increasing emissions by increasing C substrate availability for methanogens to consume CH4, and that the soil context determines the direction of the effect: High-yielding rice cultivars promote CH4 production and CH4 emissions become more pronounced in high C soils. Clearly, effects of high-yielding cultivars persist over time.

The generality of our findings was further confirmed by the results of our meta-analysis. We can only speculate about the mechanisms underlying the mitigation effect of high-yielding cultivars on CH4 emissions in our meta-analysis. Indeed, high-yielding rice cultivars differ from low-yielding cultivars in many different ways that could potentially affect CH4 emissions. For instance, compared to low-yielding cultivars, high-yielding cultivars have been shown to increase allocation to panicles (Jiang et al., 2016; Richards, 2000), and to differ in plant growth parameters (Gogoi, Baruah, & Gupta, 2008), root exudation (Lu, Wassmann, Neue, & Huang, 2000), and root oxidation activities (Zhang, Xue, Wang, Yang, & Zhang, 2009). However, our own data show that high-yielding cultivars increased root porosity, root biomass, and methanotrophic activity across multiple independent experiments. These data suggest that the effect of high-yielding cultivars on O2 transport may be general, occurring under a wide range of environmental conditions and explaining the pattern found across the experiments synthesized in the meta-analysis.

In our extrapolation, we estimated the effects of a further 10% increase in plant biomass. This represents a realistic scenario: Plant breeding efforts have increased the biomass of super rice cultivars in China by about 25% from 2000 to 2015 and are expected to increase a further 10% by 2020 (Peng et al., 2008; Yuan, 2015). In absolute terms, the reduction in the CH4 emissions caused by rice cultivars with high yield in high organic C soils was an order of magnitude larger than the emission increment in the low organic C soils (Table S1). Moreover, organic C of China’s paddy soils has increased by 7.5% from 1979–1982 to 2007–2008 (Yan, Cai, Wang, & Smith, 2011) and will likely continue to increase due to the increasingly common management practice of crop straw incorporation (Liu et al., 2014; Singh et al., 2008). Thus, our estimate of a 7.1% reduction in CH4 emissions due to high-yielding cultivars is conservative, and the real effect may be larger.

Our findings suggest that by switching to high-yielding cultivars, CH4 emissions from rice agriculture can be reduced substantially.

### FIGURE 4

Results of a meta-analysis on the effect of high-biomass rice cultivars on CH4 emissions under different soil organic C contents. Results are based on 33, 25, and 35 observations for the soil organic C ≤ 8 g kg⁻¹, 8–12 g kg⁻¹, and >12.0 g kg⁻¹ class, respectively. Error bars indicate 95% confidence intervals. The effect of high-yielding cultivars on CH4 emissions differed significantly between experimental classes (p = .0002)

The data support our hypothesis that high-yielding 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 high-yielding rice cultivars on methanotrophs and methanogens persist throughout the growing season.

Maintaining food security in the face of population growth and climate change is one of great challenges facing mankind today (Alexandratos & Bruinsma, 2012). Food security can be enhanced
through agricultural intensification, but measures that increase crop yields often increase greenhouse gas emissions too (Tilman, 1999). Here, we show that agricultural intensification can go hand in hand with greenhouse gas mitigation. Other mitigation practices advocated to curb CH$_4$ emissions from rice paddies include mid-season drainage, intermittent irrigation, no-till, and the use of alternative fertilizers (Hussain et al., 2015; Linquist et al., 2015; Zhao et al., 2016). However, these practices can result in yield losses (Pittelkow et al., 2015), are labor-intensive, and their applicability varies among rice cropping systems and countries (Bodélier, 2015). In contrast, rice cultivar improvement may be a win-win strategy, as it simultaneously decreases CH$_4$ emissions and increases grain yield. Although seeds of higher yielding cultivars will be more expensive, farmers benefit where an increase in grain yield exceeds extra cost and society benefits through the reduction in greenhouse gases. Considering the dominance of small households in most rice production areas (Zhang et al., 2016), the use of high-yielding cultivars may therefore be accepted sooner and implemented more efficiently than other mitigation practices. Along with other mitigation efforts, future policy measures aimed at reducing CH$_4$ emissions from rice cultivation should consider the use of high-yielding cultivars.

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REFERENCES


SUPPORTING INFORMATION

Additional Supporting Information may be found online in the supporting information tab for this article.

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