

1 **Modelling greenhouse gas emissions and mitigation potentials in fertilized paddy rice**
2 **fields in Bangladesh**

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22 **Abstract**

23 Emissions of greenhouse gases (GHG) from paddy rice are significant, so reducing these
24 emissions has significant potential for climate change mitigation. We investigated alternate
25 wetting and drying (AWD) as part of an integrated management approach to enhance
26 mitigation, together with combinations of mineral nitrogen (N), reduced tillage, a suitable
27 combination of plant residues and well decomposed manure. To quantify GHG emissions, and
28 the potential for mitigation without yield decline, a process-based model, DayCent, was used
29 to simulate methane (CH₄) and nitrous oxide (N₂O) emissions from paddy rice (*Oryza sativa*
30 L.) in Bangladesh. The four test sites selected were amended with mineral N fertilizer or an
31 organic amendment (rice straw). A good agreement ($p < 0.05$) was observed between model
32 simulated and measured daily CH₄ flux at most of these test sites with no significant bias. The
33 seasonal CH₄ emission from a site receiving mineral N fertilizer at a rate of 110 kg N ha⁻¹ was
34 predicted by the model to be 210 and 150 kg ha⁻¹ for the water management scenarios of
35 continuous flood (CF) and AWD, respectively. These values compare well with estimates of
36 CH₄ emissions using Intergovernmental Panel on Climate Change tier 1 methods for the
37 different water regimes. Our model results suggest emission factors for N₂O of 0.4% and 0.6%
38 of applied fertilizer under CF and AWD water regimes, respectively. Based on modelling
39 studies, AWD was found to be an important strategy not only with respect to reducing GHG
40 emissions, but also in terms of cost effectiveness. We also found that integrated management
41 is a promising option for farmers and policy makers interested in either yield increase, GHG
42 mitigation or both. Yield scaled emissions intensity under AWD was found to be about 24%
43 lower than under CF, followed by integrated management.

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45 **Keywords:** Greenhouse gas; paddy soil; water management; mitigation potential; Bangladesh.

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47 **1. Introduction**

48 The emission of greenhouse gases (GHG) from agriculture is of great environmental
49 concern, with agriculture emitting around 4.6 Gt carbon dioxide equivalent (CO₂-eq.) yr⁻¹ in
50 2010 (Tubiello et al, 2013). Of the non-CO₂ GHGs, methane (CH₄) and nitrous oxide (N₂O)
51 are the most important gases emitted from agricultural activities, with 50% and 60%,
52 respectively, of total anthropogenic emissions. In contrast to developed countries, the
53 contribution of GHG emissions from developing countries account for three quarters of total
54 global GHG emissions from agriculture (Smith, 2012). Among the agricultural sources,
55 wetland rice (*Oryza sativa* L.) production is a major contributor to the global budget of GHG
56 emissions from agriculture, which comprise 55% of global agricultural GHG emissions, of
57 which 90% is emitted in Asia (Stocker, 2013). Methane and N₂O emissions are the potent
58 GHGs that emitted from rice cultivation (Tian et al., 2018).

59 Annually, approximately 34 million tonnes (Mt) of rice (7% of world rice production)
60 are produced in Bangladesh, covering over 70% of total land (BBS, 2016). Production is
61 expected to increase by 50% to meet the demand of an increased population with changing
62 dietary preferences by 2050 (BBS, 2016). Agriculture is estimated to be one of the largest
63 sources of GHG emissions in Bangladesh, estimated at 78 Teragram (Tg) carbon di-oxide
64 (CO₂)-eq. in 2016, to which rice cultivation contributes approximately 30% of total GHG (CO₂-
65 eq.) emitted from agriculture (FAOSTAT, 2018). Although the contribution to global GHG
66 emissions from agriculture is 8-9 times lower than the other major rice producing countries
67 such as India and China, the per capita emissions in Bangladesh are essentially the same as for
68 those two countries (FAOSTAT, 2018). Concurrently, Bangladesh is recognised as one of the
69 world's most vulnerable countries to climate change, due to socio-economic conditions and its
70 geographical location (Islam and Nursey-Bray, 2017). It is necessary to focus on the climate
71 change vulnerability Bangladesh faced for the need of mitigation, and thus mitigation policy

72 in agriculture should be developed. However, In Bangladesh, emphasis has been given to
73 adaptation rather than mitigation, although there is potential to reduce GHG emissions from
74 agriculture. Resources are being invested in sectors other than agriculture, due to lack of
75 specific information to assess “business as usual” conditions (Jilani et al., 2015). For instance,
76 in the Nationally Determined Contribution (NDC) 2015, Bangladesh pledged to reduce
77 emissions from different non-agricultural sectors including power, transport and industry-
78 unconditionally by 5% and conditionally by 15% of total emissions from business as usual
79 level by 2030 (Begum et al., 2018a; Jilani et al., 2015). The detailed information on individual
80 contributions of GHG emissions for CH₄ and N₂O, considering current agricultural practices
81 are scarce. So current baseline emissions for CH₄ and N₂O, necessary to determine the
82 mitigation potential, are not yet well characterised.

83 Irrigated land in Bangladesh occupies around 60% of the total agricultural land and
84 more than half of that area is used for dry season rice, which is irrigated rice (locally known as
85 *boro*) production (BBS, 2016). For high productivity, the irrigated area needs to be expanded
86 to produce more rice for the increasing population; consequently, CH₄ and N₂O emissions are
87 expected to increase above current levels (Ali et al., 2013). Sometimes, agronomic practices
88 have opposite effects on CH₄ and N₂O emissions during the rice-growing season. For instance,
89 changing water status from continuous flooding to alternate wet and drying conditions leads to
90 a reduction in CH₄ emissions while increasing N₂O emissions and *vice versa* (Cheng et al.,
91 2013).. However, there is potential to reduce GHG emissions from paddy rice soils by
92 management of water, nutrients and other traditional practices (Smith, 2012). Beach et al.
93 (2015) found substantial GHG mitigation potential in Asia and the potential is higher for rice
94 than upland crops. Therefore, it is important to address the effect of management on both CH₄
95 and N₂O emissions to propose effective mitigation management for paddy land in Bangladesh.

96 In Bangladesh, very few field experimental studies have been conducted on GHG
97 emissions, and those that exist have investigated only one gas, either CH₄ or N₂O. A large
98 number of factors influence regional and inter-annual variability in CH₄ flux (Babu et al., 2006)
99 and empirical models are often regarded as too simple (Bell et al., 2012). Therefore, process-
100 based models are a useful supplementary method for studying GHG emissions and mitigation
101 potential under different agricultural management practices. Several modelling studies have
102 simulated SOC change and mitigation potential with rice-based cropping systems in different
103 regions of Asia (Xu et al., 2011; Bhattacharyya et al., 2007). For this study, we selected the
104 DayCent model (Parton et al., 1998) which has recently had a methanogenesis sub-model
105 added. Early studies with this version showed adequate model performance in simulating trends
106 in SOC content and CH₄ fluxes in agricultural regions – in China (Cheng et al., 2013, 2014)
107 and Brazil (Weiler et al., 2018) The DayCent was also applied in Bangladesh rice croplands to
108 determine SOC sequestration potential at site level (Begum et al., 2018b) and the GHG
109 mitigation potential in regional scale (Begum et al., 2018a), The present study aims to estimate
110 CH₄ and N₂O emissions for paddy rice in Bangladesh for nitrogen (N) fertilized study sites,
111 using the DayCent model to simulate emissions under different mitigation scenarios relative to
112 current practices. Model based CH₄ and N₂O emissions in a single paddy rice system were
113 determined and compared with estimated emissions using the Intergovernmental Panel on
114 Climate Change (IPCC) tier 1 methods for CH₄ (Lasco et al., 2006) and N₂O emissions (De
115 Klein et al., 2006).

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117 **2. Materials and methods**

118 *2.1. Site description*

119 Four experimental sites were selected which are located in the same administrative unit
120 (district) of Mymensingh. Three experiments were conducted at Bangladesh Agricultural

121 University (BAU) (site 1, site 2 and site 4), and another experimental site was located in the
122 upazilla (sub district) of Bhaluka (site 3). The test sites are at 24.75⁰ N latitude and 90.50⁰ E
123 longitude with an elevation of 18 m above sea level (Ali et al., 2014).

124 Generally, rice-rice or rice-wheat is the dominant cropping system in Bangladesh, the
125 test sites however were fallow after the rice growing season. In all sites, irrigated rice was
126 planted each year in winter (January) and harvested in summer (May). The experimental year
127 for site 1, site 2 and site 3 is 2010. Analysis for site 4 was two years-2011 and 2012. The total
128 duration of the crop was 120-140 days. Three week old seedlings of a high yielding variety
129 were transplanted in the experimental sites. The weather data from the meteorological station
130 of Mymensingh were used to drive the model for all sites (Fig. 1). The weather station was
131 located 400 m away from the site (BMD, 2016). Average temperature during the experimental
132 period (three years) recorded by Bangladesh Meteorological Department was 25.8 ⁰C. Annual
133 precipitation was 2000 mm, with 80% of rainfall received between May and September.
134 Temperature was below 20 ⁰C during the months of December-January, while the maximum
135 temperature was up to 30 ⁰C at the beginning of April until the end of August (BMD, 2016).
136 The ambient air temperature during the sampling period was 25-35 ⁰C (Ali et al., 2014).

137

138 Fig. 1.

139 2.2. Treatments and CH₄ and N₂O emissions data

140 Treatments for three sites (site 1 to 3) involved the application of N fertilizer at a rate
141 of 110-115 kg ha⁻¹ and at site 4, a combination of N fertilizer and rice straw (N+RS) was applied
142 (Table 1). The N was applied in three split applications, with around 40% of the N applied in
143 one application before transplanting. The remaining portion was applied as two equal split
144 applications at tiller initiation stage (about three weeks after transplantation) and panicle

145 initiation stage (about six weeks after rice transplantation) (Ali et al., 2013). Rice straw alone
146 was used as an organic amendment and was applied before rice cultivation at site 4. Each
147 treatment had three replications. The soil details and properties are summarized in Table 1.
148 Details of the measurements are described in Ali et al. (2012, 2013 and 2014).

149 Data for CH₄ and N₂O were available for CF and AWD water regimes for site 1 while
150 for the other sites, only CH₄ emissions under CF conditions were measured. In the CF
151 condition, the soil was fully saturated for the entire crop growing season, water level in the rice
152 field was kept at 5cm depth while under AWD systems, the rice field was irrigated during the
153 final land preparation to rice planting time, active tillering stage and flowering stages. The field
154 was kept moist during the rest of the period (Ali et al., 2013). Static closed chambers were used
155 for gas sampling during rice cultivation. The air gas samples from the transparent glass
156 chamber (diameter 60 cm, and height 110 cm) were collected by using 60-ml gas-tight syringes
157 at 0, 15 and 30-minute intervals after chamber placement over the rice-planted plots. The
158 surface area of each chamber was 0.25 m² (0.5X0.5 m²). While gas sampling, the chamber was
159 placed over six hills of rice vegetation. There were four holes at the bottom of each chamber
160 through which water movement was controlled. Gas samples were simultaneously analysed
161 with a modified gas chromatograph equipped with a flame ionization detector and an electron
162 capture detector (Wang and Wang, 2003). The detailed description of sample analysis was
163 found in Ali et al., (2012, 2013). Methane and N₂O emissions from paddy fields were calculated
164 by using the equation (Rolston, 1986):

$$165 \quad F = \rho * V/A * \Delta c/\Delta t * 273/T \quad (1)$$

166 where, F = CH₄ flux (mg m⁻² hr⁻¹) or N₂O flux, ρ = gas density (0.714 mg cm⁻³), V = volume
167 of chamber (m³), A = surface area of chamber (m²), Δc/Δt = rate of changes in CH₄ or N₂O gas
168 concentrations in the Chamber (mg m⁻³ hr⁻¹), and T (absolute temperature) = 273 + mean

169 temperature in chamber (⁰C) (Ali et al., 2013). Gas samples were collected 2 times per day,
170 once a week during the cropping season. On average 7-8 observations were obtained for CH₄
171 emissions and 7 values for N₂O (Ali et al., 2012; Ali et al., 2013; Ali et al., 2014).

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192 **Table 1**

193 Initial physico-chemical characteristics of four N fertilized test sites in Bangladesh.

Location (Experimental year)	Treatment	Texture	SOM (%)	Water regime	BD (g cm ⁻³)	pH	Available measured data	Ref
BAU site 1 (2010)	110 kg N ha ⁻¹	Silt loam	2	CF, AWD	1.18	6.2	CH ₄ N ₂ O	Ali et al., 2013
BAU site 2 (2010)	115 kg N ha ⁻¹	Silt loam	2.1	CF	1.25	6.1	CH ₄	Ali et al., 2012
Bhaluka site 3 (2010)	115 Kg N ha ⁻¹	Silty Clay loam	2.3	CF	1.29	5.8	CH ₄	Ali et al., 2012
BAU site 4 (2011-2012)	110 kg N ha ⁻¹ + 2 t ha ⁻¹ rice straw (total C and N 39.50% and 0.95% respectively)	Clay loam	1.78	CF	1.34	5.9	CH ₄	Ali et al., 2014

194 BAU: Bangladesh Agricultural University, SOM: Soil organic matter, CF: Continuous flood, AWD: Alternate wet and drying,

195 BD: Bulk density, Ref: References

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199 2.3. *Model description and simulations*

200 We used the most recent version of the DayCent ecosystem model (Parton et al., 1998),
201 developed for paddy rice (Cheng et al., 2013). It is the daily time-step version of the
202 CENTURY model, and provides daily outputs of net primary production and heterotrophic
203 respiration. The model simulates biogeochemical processes associated with carbon, N,
204 phosphorus, and sulphur cycling, including SOM decomposition, nitrification and
205 denitrification, plant production and soil water dynamics, and, in the version used in this study,
206 methanogenesis (Cheng et al., 2013; Hartmann et al., 2016). The methanogenesis sub-module
207 simulates CH₄ production based on C substrate supply derived from decomposition of SOM
208 and root rhizodeposition. Soil texture, soil pH, redox potential (Eh), soil temperature, climate
209 and agricultural management impact on methanogenesis, thereby CH₄ formation (Cheng et al.,
210 2013). DayCent does not simulate diffusion of CH₄ through the surface water to the atmosphere
211 because it is considered a minor pathway for CH₄ emissions (Cheng et al., 2013). Ebullition
212 occurs when the soil CH₄ concentration exceeds a critical state that leads to formation of
213 bubbles (Cheng et al., 2013; Hartmann et al., 2016). The trace gas sub-model of DayCent
214 simulates soil N₂O and NO_x gas emissions from nitrification and denitrification processes.
215 Daily denitrification rates are estimated for each soil layer based on nitrate (NO₃⁻)
216 concentration, heterotrophic respiration (as a proxy for labile C availability), water content,
217 texture, and temperature (Del Grosso et al., 2008). Detailed information about model concepts
218 and mechanisms is described in greater detail elsewhere (Del Grosso et al., 2008; Cheng et al.,
219 2013; Hartmann et al., 2016). The DayCent model has been applied for different land uses,
220 including grasslands (Parton et al., 1998), agricultural lands (Begum et al., 2017, Senapati et
221 al., 2016), forests (Cameron et al., 2013), and savannas (Parton et al., 1993).

222 DayCent requires precipitation, and maximum and minimum temperature at daily time
223 steps, which is based on a meteorological station at Mymensingh for this study. Based on

224 available SOM data measured to a 15 cm depth, the initial SOC stock (in t ha⁻¹) was estimated
 225 using an equation in Nayak et al. (2015), multiplying measured %SOM by 0.58, depth (in cm)
 226 and BD (in g cm⁻³) which was used in a study of Chinese croplands including 50 studies of rice
 227 ecosystems. The simulated DayCent values, which are estimated for 20 cm, were adjusted to
 228 15 cm depth by dividing DayCent outputs by 1.33. Field capacity (FC), wilting point (WP) and
 229 saturated hydraulic conductivity were estimated using a pedo-transfer function of Saxton and
 230 Rawls (2006). SOC pools in the model were initialized with a model spin-up for 1500 years
 231 using native vegetation and historical agricultural management as suggested by previous
 232 applications of the model (Begum et al., 2018b; Cheng et al., 2013; Del Grosso et al., 2006).

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234 2.4. Statistical methods

235 Performance of the model was evaluated with statistical routines provided in
 236 MODEVAL (Smith et al., 1997; Smith and Smith, 2007). The sample correlation coefficient
 237 (r) was used (equation 2) to test for association between the modelled and measured values
 238 over time. Modelled and measured daily flux of CH₄ and N₂O emissions were compared by
 239 calculating the root mean square error ($RMSE$, equation 3), which indicates total difference
 240 between observed and predicted values (Smith et al., 1997). The mean difference between
 241 observation and simulation (M) was calculated to assess bias in the modelled results (equation
 242 4).

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$$244 \quad r = \frac{\sum_{i=1}^n (O_i - \bar{O})(P_i - \bar{P})}{\sqrt{[\sum_{i=1}^n (O_i - \bar{O})^2][\sum_{i=1}^n (P_i - \bar{P})^2]}} \quad (2)$$

$$245 \quad RMSE = \frac{100}{\bar{O}} \sqrt{\frac{\sum_{i=1}^n (P_i - O_i)^2}{n}} \quad (3)$$

$$246 \quad M = \frac{\sum_{i=1}^n (O_i - P_i)}{n} \quad (4)$$

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248 Where \bar{O} and \bar{P} are the mean values of observed and predicted data, respectively, and O_i and
249 P_i indicate the observed and predicted values at the i th iteration, respectively, and n is the
250 number of samples. The significance of r and M were tested using an F-test (at probability
251 levels of $p = 0.05, 0.01$ and 0.001), and a Student's two-tailed t-test (critical at 2.5%).

252 2.5. Mitigation scenarios and net GHG emissions

253 The model was used to simulate the impact of alternative management practices on
254 mitigation of GHG emissions. Management associated with AWD, as practised in test site 1,
255 was also included here as a mitigation option to estimate total GHG emissions, including
256 predicting CO₂ emissions under this management. Along with this single-practice mitigation
257 scenario, two integrated approaches were tested considering tillage, residue management, N
258 fertilizer, and two different types of manure. The full list of practices considered was:

- 259 • RT: use of reduced tillage (RT, sowing with less disturbance to the top soil) instead of
260 conventional tillage (CT).
- 261 • Rsd20: 20% of straw removal instead of the baseline of 5%.
- 262 • CD: well decomposed cowdung (CD) of approximately 8 t ha⁻¹ (substitution of N in
263 CD for baseline mineral fertilizer N) with 1.33% N and C:N ratio of 31.50 (Ali et al.,
264 2014).
- 265 • GM: well decomposed green manure (GM) of approximately 4 t ha⁻¹ (substitution of N
266 in GM for baseline mineral fertilizer N). *Sesbania (Sesbania rostrata)* biomass with
267 2.80% N and C:N ratio of 23.50 was considered as green manure (Ali et al., 2014).
- 268 • AWD
- 269 • IM1: Integrated management of RT with residue return of 15% (Rsd15), CD with
270 substitution of 60% baseline mineral N fertilizer, AWD and mineral N fertilizer at a
271 current rate of 110 kg ha⁻¹.

- 272 • IM2: All management was same as in IM1 except manure was replaced with a GM with
273 40% baseline N substitution.

274 Default model parameters were used to simulate plant production and different tillage
275 intensities, as presented in Table 2 (Hartmann et al., 2016).

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292 **Table 2**

293 The plant production and cultivation parameter file of DayCent model used for the current study to simulate CH₄

294 and N₂O emissions from rice cropland, Bangladesh

Name of the file	Parameter	Description	Unit	Value
Crop.100	PRDX	Coefficient for calculating potential aboveground monthly production as a function of solar radiation outside the atmosphere	Scaling factor, (g C production) m ⁻² month ⁻¹ Langley ⁻¹	3.00
	PPDF (1)	Optimum temperature for production for parameterization of a Poisson Density Function curve to simulate temperature effect on growth	°C	25
	PPDF (2)	Maximum temperature for production for parameterization of a Poisson Density Function curve to simulate temperature effect on growth	°C	45
	HIMAX	Maximum harvest index		¹ 0.42-0.48
	TMXBIO	Maximum above ground biomass at the end of growing season	g biomass m ⁻²	¹ 1000-1100
Cult.100	CULTRA (5)	Fraction of standing dead transferred to top soil layer		CT: 0.6 RT: 0
	CLTEFF	Cultivation factor for soil organic matter decomposition; functions as a multiplier for increased decomposition in the month of the cultivation		CT: 3.85 RT: 3.41

295 ¹The ranges varies among test sites; CT: conventional tillage, RT: reduced tillage.

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303 GHG emissions (kg CO₂-eq. ha⁻¹ yr⁻¹) were estimated using global warming potential
304 (GWP) (CO₂-eq.) over a 100 year time span (Forster et al., 2007) (equation 5).

$$305 \text{ GHG} = (25 \times [\text{CH}_4]) + (298 \times [\text{N}_2\text{O}]) \quad (5)$$

306 where GHG is the total CH₄ and N₂O emissions in kg CO₂-eq. ha⁻¹ yr⁻¹. The GWP for CH₄ and
307 N₂O are 25 and 298 over a 100-year time span (Forster et al., 2007). To get relative changes of
308 GHG emissions, baseline emissions were deducted from emissions under mitigation
309 management, and then the difference divided by baseline emissions (equation 6), all expressed
310 in kg CO₂-eq. ha⁻¹ yr⁻¹.

$$311 \text{ Relative } \Delta\text{GHG} = (\text{GHG}_{\text{Miti}} - \text{GHG}_{\text{BL}})/\text{GHG}_{\text{BL}} \quad (6)$$

312 Where relative ΔGHG is the relative change of emissions associated with different
313 management options. GHG_{Miti} is emissions under the mitigation scenario, and GHG_{BL} is
314 baseline emissions. Negative values suggest an alternative scenario could mitigate GHG
315 emissions; positive values indicate an increase in GHG emissions relative to the baseline. The
316 relative changes in paddy rice yield were also calculated so that the combined impact of
317 management change on both yield and GHG emissions could be tracked (equation 7).

$$318 \text{ Relative } \Delta\text{yield} = (\text{yield}_{\text{Miti}} - \text{yield}_{\text{BL}})/\text{yield}_{\text{BL}} \quad (7)$$

319 Where relative Δyield is the relative changes of yield associated with different mitigation
320 options. $\text{yield}_{\text{Miti}}$ is crop yield under the mitigation scenario, and GHG_{BL} is baseline crop yield,
321 all expressed in kg ha⁻¹ yr⁻¹. Additionally, to determine the emissions intensity of production,
322 GHG emissions per unit of crop yield were calculated (equation 8).

$$323 \text{ GHGI} = \text{GHG}/\text{yield} \quad (8)$$

324 where GHGI is the GHG emission intensity (kg CO₂-eq. kg⁻¹ yield), GHG is the total emissions
325 (CH₄ and N₂O) (kg CO₂-eq. ha⁻¹ yr⁻¹) and yield denotes crop production (kg⁻¹ yr⁻¹). The relative

326 changes of GHGI under different mitigation option to that of baseline were calculated to
327 determine net mitigation potential of the selected management (equation 9).

$$328 \text{ Relative } \Delta\text{GHGI} = (\text{GHGI}_{\text{Miti}} - \text{GHGI}_{\text{BL}}) / \text{GHGI}_{\text{BL}} \quad (9)$$

329 where $\text{GHGI}_{\text{Miti}}$ and GHGI_{BL} denotes GHG emission intensities under mitigation and baseline
330 management respectively.

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332 **3. Results**

333 *3.1. Simulated CH₄ and N₂O emissions*

334 The observed and simulated daily CH₄ fluxes for four experimental sites are presented
335 in Fig. 2 a-f. Daily CH₄ flux for all experimental sites increased from the tillering stage, with
336 the highest peak observed at the flowering to maturity stage (77-100 day after transplantation),
337 during the month of March-April, and gradually declined towards the harvesting stage.
338 DayCent simulates the dynamics of the observations quite well except for one instance at BAU
339 site 4 for the year 2012 ($p < 0.05$) (Table 3). Overall, the daily CH₄ flux was lower in AWD
340 and higher under combined application treatments (N+RS) than the emissions under CF with
341 mineral N application, observed both in simulations and observations. Although the statistical
342 error between modelled and measured was from 0.01-0.08 g m² d⁻¹ (RMSE 25-53%), no bias
343 was observed at either of the test sites (Table 3). In contrast to the daily fluxes, the cumulative
344 seasonal CH₄ emissions are over estimated compared to the reported values for CH₄ emissions
345 (Table 4). However, the modelled seasonal CH₄ emissions under different management and
346 treatment follows a similar trend as seen in the observations, with the following order from
347 lowest to highest emissions: N treatment with AWD > N treatment with CF > N+RS treatment
348 with CF. Compared to CF, AWD reduced CH₄ emissions by nearly 30%, according to both
349 observations and simulations. On average, seasonal CH₄ emissions increased by around 6%

350 with combined treatment of N+RS application under CF (Site 4) compared to mineral N-only
351 application (site 1), while the model simulated an increase of nearly 17%. The IPCC estimated
352 values with N fertilized paddy field under CF and AWD, and with rice straw application
353 suggest average seasonal CH₄ emissions of 200, 114 and 224 kg ha⁻¹ which is close to the
354 values predicted by DayCent under similar management (Table 4).

355 The model simulated a relatively larger peak in daily N₂O flux under CF conditions
356 after the third fertilizer application. The emissions tend to be lower after fertilization, until the
357 land is drained (before harvesting) (Fig. 2g). However, the trend of N₂O emissions was
358 underestimated by the model (0.15 g m⁻² d⁻¹) during the entire cropping seasons without bias
359 (Fig. 2g, Table 3). The maximum peak of daily N₂O flux was also found under AWD treatment,
360 both in observations and simulations, showing peaks three times higher than under the CF
361 treatment. A few large peaks observed in the field before harvesting were not captured by the
362 model. Although there was not close agreement between modelled and measured flux ($p <$
363 0.05 , $RMSE = 67\%$), no systematic bias was found for either of the management types (Table
364 3). Overall seasonal N₂O emissions were simulated as 0.61 kg ha⁻¹ by model, but were observed
365 to be 0.98 kg ha⁻¹ in the measurements (Table 4). N₂O emissions measured in the test site under
366 AWD were 78% higher than measured under CF, while the model predicts 36% higher N₂O
367 emissions under AWD conditions to that of simulated under CF. Based on mineral fertilizer
368 applied, IPCC estimated values in a lowland paddy soil irrespective of water management was
369 0.52 kg ha⁻¹. Our DayCent values were 13% lower under CF management and 17% higher
370 under AWD than those IPCC estimated values.

371 The average crop yield under different management types varied from 4240 to 5070 kg
372 ha⁻¹ and as with the observations, a higher yield (14%) was attained by the model with
373 combined application of N+RS (Site 4, Table 4).

374

375 **Fig. 2.**

376 **Table 3**

377 The calculation of *r*, *RMSE* and *M* showing *F* (P=0.05, 0.01, 0.001) and critical *t* (2.5% Two-tailed) between
 378 simulated and observed daily CH₄ and N₂O emissions under CF and AWD water management at the two paddy
 379 rice test sites (description of individual site tests is available in Table 1).

Location (Experimental year)	Treatment	Water regime	¹ Available measured data	<i>r</i>	<i>RMSE</i> (%)	<i>M</i> g m ⁻² d ⁻¹ / mg m ⁻² d ⁻¹
BAU site 1 (2010)	110 kg N ha ⁻¹	CF	CH ₄ (8)	0.82**	44.86	0.04 ^{ns}
		AWD	CH ₄ (8)	0.80*	52.78	0.01 ^{ns}
		CF	N ₂ O (7)	0.10 ^{ns}	49.90	0.15 ^{ns}
		AWD	N ₂ O (7)	0.65 ^{ns}	67.35	0.55 ^{ns}
BAU site 2 (2010)	115 Kg N ha ⁻¹	CF	CH ₄ (8)	0.96***	25.35	0.01 ^{ns}
Bhaluka site 3 (2010)	115 Kg N ha ⁻¹	CF	CH ₄ (7)	0.73*	42.98	0.05 ^{ns}
BAU site 4 (2011)	110 kg N ha ⁻¹ + 2t ha ⁻¹ rice straw (N+RS)	CF	CH ₄ (8)	0.75*	39.22	0.02 ^{ns}
BAU site 4 (2012)	110 kg N ha ⁻¹ + 2t ha ⁻¹ rice straw (N+RS)	CF	CH ₄ (8)	0.71 ^{ns}	42.65	0.08 ^{ns}

380 ¹Figure in parenthesis in column 4 denotes sample number

381 *Significant correlation (*r*) between modelled and measured values at p <0.05, or significance mean error (*M*) at p = 0.025.

382 ** Significant correlation (*r*) between modelled and measured values at p <0.01.

383 *** Significant correlation (*r*) between modelled and measured values at p <0.001.

384 ns = non-significant between modelled and measured values at p <0.05, or no significance mean error (*M*) at p = 0.025.

385 **Table 4**

386 Yearly observed and simulated CH₄ (four sites), N₂O emissions (first site) along with IPCC default values and
 387 crop yield under contrasting water and nutrient management on selected sites (description of individual site tests
 388 are available in Table 1).

389 NA: Measured data not available

Test site	Water regime	CH ₄ (kg ha ⁻¹ yr ⁻¹)			N ₂ O (kg ha ⁻¹ yr ⁻¹)			Crop yield (kg ha ⁻¹ yr ⁻¹)	
		Measured	Modelled	IPCC	Measured	Modelled	IPCC	Measured	Modelled
BAU site 1 (2010)	CF	124	210	190	0.55	0.45	0.52	4290	4241
BAU site 1 (2010)	AWD	90	150	114	0.98	0.61	0.52	4350	4118
BAU site 2 (2010)	CF	106	226	206	NA	NA		4189	4593
Bhaluka site 3 (2010)	CF	129	200	200	NA	NA		4450	4980
BAU site 4 (2011)	CF	125	246	224	NA	NA		4900	5070
BAU site 4 (2012)	CF	140	251	224	NA	NA		5020	5050

390

391

392 3.2. Modelling GHG mitigation

393 Changes in management for GHG mitigation in most cases lead to opposite impacts on
394 CH₄ and N₂O emissions (Fig. 3a). The two exceptions are residue management and RT, which
395 show hardly any change (up to 2%). Application of manure in place of mineral N fertilizer
396 reduces N₂O emissions up to nearly 50% (with CD application) while it increases CH₄
397 emissions by nearly same amount (with GM application) compared to the baseline. The
398 opposite trend was seen for other management options, including an increase in N₂O emissions
399 by up to 70% under integrated management along with GM (IM2), and up to a 30% decrease
400 in CH₄ emissions under AWD management (site 1 test simulations).

401 Comparing the relative changes between net GHG emissions (CO₂-eq. ha⁻¹ yr⁻¹) using
402 GWP for a 100 year time horizon and yield, GHG emissions were lower with AWD water
403 regimes by 26%, with a negligible yield decline (2%) (Fig. 3b). GHG emissions increased by
404 up to 50% under single manure application, which also reduced yield by around 34-55%. Based
405 on the model results, three options can be selected for reducing emissions without having a
406 negative impact on yield, including from highest to lowest as: AWD > IM1 > IM2 > RT while
407 the best outcomes are achieved under integrated management (IM1 and IM2) which reduced
408 GHG emissions by up to 6% (with IM1), and also increased yield by up to 6% (with IM2).

409

410 **Fig. 3.**

411 Maximum GHG reductions were seen for AWD, with a yield scaled emissions intensity
412 about 24% lower than under CF, followed by IM1 and IM2, respectively (Fig. 4). The change
413 in emissions intensity was negligible under adoption of tillage and residue management (<3%),
414 while it was predicted to be 1.3-1.5 times higher under manure application scenarios.

415

416 **Fig. 4.**

417

418 **4. Discussion**

419 *4.1. Modelled CH₄ and N₂O emissions and yield*

420 Simulation of substrate C available for methanogenesis by DayCent under different
421 water and nutrient management is crucial for predicting CH₄ emissions accurately (Cheng et
422 al., 2013). A large CH₄ flux was simulated at plant maturity stage in the month of April-May,
423 when carbohydrates derived from plant was greater. Higher temperature is another controlling
424 factor that favours methanogenic bacteria, hence CH₄ emissions (Zhang et al., 2013; Neue and
425 Scharpenseel, 1984). In the test sites, higher temperatures were observed at the plant maturity
426 stage, which favours methanogenic activity (Ali et al., 2012). In response to measured soil
427 temperature of 26-32 °C, the simulated soil temperature was predicted to be 20-30 °C.
428 DayCent-simulated soil Eh under CF water regime was relatively high, predicted to be -188
429 mV compared to that of -81 mV under AWD conditions. The measured Eh in the real field
430 under CF and AWD water regime were reported as -95 mV and -71 mV, respectively (Ali et
431 al., 2013).

432 A difference between seasonal modelled and measured CH₄ emissions was observed,
433 but this might be expected since cumulative emissions were calculated using relatively few
434 data points (Ali et al., 2013; Ali et al., 2014). The impact of different nutrient management and
435 water regimes on CH₄ emission was satisfactorily replicated by the model. Compared to CF,
436 DayCent simulated lower water filled pore space, enhanced aerobic microbial activity and
437 thereby Eh, and overall reduced CH₄ emissions under AWD conditions. In contrast, increasing
438 labile C with organic matter application (rice straw for site 4) in a continuously flooded soil
439 tended to increase CH₄ emissions compared to a mineral N fertilized sites (site 1). The
440 cumulative seasonal CH₄ emissions for irrigated rice from Bangladesh field experimental

441 studies, found to vary from 98 to 800 kg ha⁻¹ depends on water management, nutrient
442 management and farming practices (Ali et al., 2013; Ali et al., 2014, Frei et al., 2007). Using
443 an empirical model CH₄MOD2.5, the average annual CH₄ emission from irrigated rice with
444 mineral N and farmyard manure application was estimated by Khan and Saleh (2015) to be 237
445 kg ha⁻¹. Modelled seasonal CH₄ emissions compared well with estimates using the IPCC Tier
446 1 methodology and previous studies.

447 Our model results showed that N₂O emissions peaks were driven by water management
448 and fertilization. Both the observations and DayCent simulations suggest lower N₂O emissions
449 for flooded paddy soil compared to AWD management. A slight underestimation of N₂O
450 emissions by the model in CF conditions could be attributed to a limited source of N, or lack
451 of nitrification under flooded conditions. The default values of N₂/N₂O ratio in DayCent were
452 set in a way to simulate less N₂O emissions from saturated soils, while in real fields there might
453 be external sources of N, including from aquatic weeds and algae (Roger and Ladha, 1992;
454 Ladha et al., 2016); these are not considered in the N and C balance of the model. Further, O₂
455 released from the rhizosphere zone of paddy fields might enhance nitrification and
456 denitrification processes and increase N₂O emissions (Babu et al., 2006). Although there are
457 no zero input treatments among the selected tested sites, Gaihre et al., (2015) found around
458 0.07 kg ha⁻¹ N₂O emissions from unfertilized plots at two irrigated rice test sites associated
459 with CF conditions in Bangladesh. This is one potential reason for the slight underestimation
460 of modelled N₂O emissions compared to the measurements. Relatively higher N₂O emissions
461 were simulated under AWD management compared to CF, which could be attributed to
462 anaerobic-aerobic conditions that influence microbial nitrification, thereby the denitrification
463 process. In AWD systems, the model simulates enhanced nitrification in presence of O₂, and
464 denitrification when the soil is saturated. In reality, it is not always possible to control the water
465 level in paddy fields. The measured peak at the pre-harvesting stage missed by the model. The

466 soil NO_3^- -N concentration in the tested site was found to be three times higher in AWD systems
467 compared to CF measured at pre-harvesting stage, (not shown), while similar N_2O emissions
468 were predicted by the model for the same period. The emission factor (EF) for N_2O emissions
469 under flooded paddy rice was simulated by the model to be 0.4% of applied fertilizer, which is
470 slightly lower than the observations (0.5% of applied fertilizer), but slightly higher than the
471 IPCC default EF (0.3% of applied fertilizer). A relatively higher EF for AWD systems (0.6%
472 of applied fertilizer) suggests that a separate EF for paddy rice under alternative water
473 management should be considered, as was suggested by Shepherd et al. (2015). Their study
474 found EF values (relative to N applied) for paddy rice under urea application in neutral soil of
475 0.03% under CF, and 0.31-0.72% under reduced water use management, and in high acidity
476 soil they found an EF of 0.16% under CF and 0.22% under intermittent saturation conditions.
477

478 *4.2. Mitigation scenarios and net GHG balance*

479 As with previous studies (Ali et al., 2013; Ma et al., 2013; Wang et al., 2013), our
480 modelled results found a trade-off among the major GHGs with different management options.
481 Based on modelled results, it is recommended that both CH_4 and N_2O need to be considered
482 together along with the yield impact before implementation of alternative management
483 practices. Methane emissions from flooded paddy rice appear to be dominant followed by N_2O
484 emissions, irrespective of management, as observed in previous studies (Zhang et al., 2013).
485 Applying the same amount of N in the form of manure does not give as high a yield, and
486 increases total GHG emissions. Our model results suggest that N mineralization through
487 application of manure might not be large enough to ensure the potential yield. More residue
488 incorporation, or use of manure, might not be possible if yields are reduced, and may be limited
489 by socioeconomic consequences in Bangladesh. Increasing levels of crop residue incorporation
490 in Bangladesh is quite challenging because of the use of residues for other household purposes,

491 e.g., as a fuel or fodder for animals (Hossain, 2001; Haider, 2013; Huq and Shoaib, 2013).
492 Similarly, there is a restriction to applying all the manure produced in Bangladesh as an organic
493 amendment, because CD has alternative uses, e.g., as fuel and biogas (BLRI, 2017; Huq and
494 Shoaib, 2013). Among the selected single scenarios, AWD management is considered to be an
495 effective option which has only a slight impact on current yield, but reduces total GHG
496 emissions by 26% relative to the baseline. This outcome agreed well with the findings Ali et
497 al. (2013) which were 24-26% reductions in total GHG emissions from AWD practices on this
498 site. The model also matches well with decrease of emission intensity of 24% reported by Ali
499 et al. (2013). Although additional costs are likely to be higher initially due to the need for
500 weeding, overall labour costs are found to decrease compared to traditional systems (Rejesus
501 et al., 2011), and water is saved (Price et al., 2013). Our model results also suggest that
502 integrated management associated with RT coupled with 15% crop residue return, application
503 of GM along with current mineral N fertilizer and AWD management appears to be the best
504 option for reducing GHG emissions and increasing crop yield. Emission intensity also found
505 to be reduce under this approach. The impact on GHG mitigation under this integrated
506 management, however, is lower than for AWD only, but positively impacts on yield. Applying
507 DNDC model in China paddy field, Tian et al., (2018) found that combined midseason drainage
508 and balanced fertilization leads to reduced CH₄ and N₂O emissions without yield penalty.

509 DayCent cannot simulate water level of the rice field but there is scope to improve the
510 water sub-model in DayCent to better reflect the real field conditions. The current version of
511 DayCent manipulated the water table by FLOD events set by the model. The Eh in soil changes
512 based on the flooded or drainage conditions. Continuous flooding, whether by rainfall,
513 irrigation or both, would be a FLOD 2 period in the model schedule, with maximum Eh of -
514 250 mV (Cheng et al., 2013, Weiler et al., 2018). The water conditions under rainfall do not
515 saturate in the model, therefore fixed values (-20 mV) were indicated as FLOD 1. Eh

516 approximations are specific to the methane model. Further development of the Eh algorithm
517 was suggested by Weiler et al., (2018), where they found contrasting results between simulated
518 and observed CH₄ emissions in flood-irrigated rice paddy fields under no tillage in southern
519 Brazil. Soil water and gas filled pore space in the current version are normal inputs to the N₂O
520 emissions, but are not currently considered in the methane Eh equations.

521 The model was tested with only 7-8 observations. The data were not recorded routinely
522 in an hourly or daily basis due to lack of funding and manpower. There are no field experiments
523 that test the efficacy of mitigation practices in Bangladesh, which is why we are attempting to
524 model them here. In this paper, we have tested the model against the best (though imperfect)
525 datasets available in Bangladesh to show that the model can adequately capture the impacts of
526 soil types, climate, management and water status on CH₄ and N₂O emissions. We aimed to
527 show through this step that the model is able to capture the influence of these factors on
528 emissions. Having demonstrated that the model performs adequately, and that we have some
529 confidence in model predictions from this validation, we have then applied the model to explore
530 potential mitigation options. Given that there are no field data on mitigation options, we cannot
531 perform a further validation of the model; instead we aimed to show direction and magnitude
532 of impacts, which we hope will be tested through future field experiments. The testing of the
533 model against the only available field data is our only option for validation, and from our
534 results, we suggest that model performance is adequate for testing the mitigation options. We
535 have focussed on relative changes in GHG emissions from different management practices
536 rather than absolute values, due to the acknowledged limitations in the validation data. We
537 hope this study can be used to guide further research on CH₄ and N₂O emissions from
538 Bangladesh paddy soils.

539

540 Due to lack of measured data, GHG mitigation is estimated here without considering
541 CO₂ emissions. The contribution of CO₂ emissions from agriculture is lower to that of other
542 anthropogenic sources (Smith et al., 2007, Cheng et al., 2014). Additionally, for paddy fields,
543 CH₄ and N₂O emissions dominate the overall GHG balance (Wang et al., 2017). For this study,
544 the relative GHGI under the selected integrated approach was predicted to be the same without
545 considering CO₂ (not presented). Total GHGI under AWD was found to be 5% lower when
546 considering all three GHGs compared to when only CH₄ and N₂O were considered. Although
547 CO₂ emissions from agriculture are small, it is crucial to estimate SOC sequestration potential
548 from paddy fields in order to improve soil quality. Around 90% of total GHG mitigation from
549 agriculture globally is estimated to be from SOC sequestration (Smith et al., 2007). Applying
550 DayCent model in a long term double rice system in Bangladesh, located at BAU, Begum et
551 al., (2018b) predicted SOC changes of -0.05 to 0.36 t C ha⁻¹ yr⁻¹ under different management
552 scenarios. Therefore, further refinement is possible to measure SOC, CH₄ and N₂O for the same
553 sites to evaluate total GHG mitigation potentials and yield impacts of GHGs in Bangladesh.

554 If these results could be scaled to the country level, if 50% of the harvested area under
555 irrigated rice were under integrated management, a reduction of approximately 1.40 Tg CO₂-
556 eq. yr⁻¹ could be realised. This rough estimate could vary depending on availability and
557 applicability of manure, and taking into account the amount already being applied. Crop yield
558 is considered as the main priority in developing countries, so there is no opportunity to reduce
559 mineral fertilizer use, but farmers have been encouraged to increase N use efficiency. Deep
560 placement of fertilizers, rather than applying urea in a traditional broadcast method, increases
561 N use efficiency and yield while reducing N₂O emissions (Gaihre et al., 2015). Changing the
562 composition in mineral fertilizer is another mitigation approach that may increase yields while
563 reducing N₂O emissions. An alternative model experiment (data not presented), using 50%
564 ammonium N (NH₄⁺-N) and 50% NO₃⁻-N, compared to 100% of NH₄⁺-N as in urea, or using

565 nitrification inhibitor under integrated approach, reduced net GHG emissions by 4%, while
566 increasing yield by 6%. Recent field experimental studies Bangladesh paddy rice field, found
567 an increase of both CH₄ but mitigate N₂O emissions with the use of biochar amendment while
568 increasing yields (Ali et al., 2013). The development of biochar amendment simulations in
569 DayCent is ongoing.

570

571 **4. Conclusion**

572 The results presented here suggest that there is scope to reduce GHG emissions from rice
573 production in Bangladesh by modifying current agricultural management practices. By
574 modifying traditional flooding practice, it is possible to reduce net GHG emissions (CO₂-eq.
575 ha⁻¹ yr⁻¹) from paddy soil by ~26%. Although such management leads to a slight yield decline,
576 farmers can also save water from irrigated rice by adoption of AWD systems. Integrated
577 management that consider RT, more residue return, AWD and GM application along with
578 mineral N fertilizer, is predicted to increase yield while reducing emissions. As farmers become
579 more interested in yield, an integrated approach is likely to be the most effective approach to
580 maintain or increase yields while also reducing GHG emissions in rice production systems of
581 Bangladesh. Further measurements of emissions for tillage and manure (CD and GM) practices
582 are necessary before implementing the model outcomes.

583

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590

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