1 <u>RESEARCH ARTICLE</u>

2 SPAD Chlorophyll Meter-Based Real-Time Nitrogen Management in

3 Manure-Amended Lowland Rice

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19 ABSTRACT

20 Precise adjustment of nitrogen (N) application time and dose under real-time nitrogen 21 management (RTNM) is crucial for optimal benefits. The study aimed to assess and recommend 22 best combination of threshold SPAD-502 chlorophyll-meter reading (SCMR) and topdressing N 23 rate for lowland flooded-rice (manure amended at 5 t ha⁻¹) to maximize yield, N use efficiency, 24 and economic gains. Four SCMR-based N scheduling (SCMR₃₄, SCMR₃₆, SCMR₃₈, SCMR₄₀), 25 each with three topdressing N rates (15, 20, 25 kg N ha⁻¹) as RTNM were evaluated and compared with fixed-time N management (FTNM) (100 kg N ha⁻¹, 4 splits). Topdressing of 20 26 27 kg N ha⁻¹ at \leq SCMR 38 (SCMR₃₈N₂₀) led to a higher yield (+8%), N use efficiency (+43%), net 28 return (+11%) and reduced N input (-18%) compared to FTNM. The yield advantage with 29 SCMR₃₈N₂₀ was attributed to higher internal N use efficiency and optimal source-sink balance 30 (higher harvest index). The best-fit treatment $SCMR_{38}N_{20}$ along with a low-yielding treatment 31 $(SCMR_{36}N_{20})$ was tested across seven rice cultivars for validation. Treatment SCMR₃₈N₂₀ led to 32 yield advantage (+7%), but significant only for four cultivars. In lowland rice, RTNM can 33 therefore increase yield and N use efficiency. However, cultivar-specific adjustments to the 34 SCMR threshold and N rate are necessary.

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36 **KEYWORDS**: Real-time N management, Lowland flooded-rice, Lower Indo-Gangetic plain,
37 Agronomic N use efficiency, N recovery efficiency, Multivariate regression analysis.

38 Introduction

Rice (*Oryza sativa* L.) is predominantly grown in lowland soils that are mostly deficient in available nitrogen (N) (Choudhury and Kennedy 2005; Hou et al. 2019). The current practice of N management [fixed-time N management (FTNM)] in rice has several drawbacks including lower N use efficiency, groundwater and atmospheric pollutions, and suboptimal productivity. Now, there is a major concern over the decline in partial N factor productivity in flooded-rice production primarily attributed to discrepancy in crop N demand and supply and higher magnitude of losses (Fagodiya et al. 2020). This has led to an interest in crop demand-driven

real-time N management options to maximize nutrient use efficiency, yield, and profitability in
rice production systems (Bandaogo et al. 2015).

48 Crop season temporal N dynamics in anoxic-flooded rice soils depends on time and rate 49 of fertilizer/manure, crop nutrient removal, and N losses, and thus, ensuring optimal N 50 bioavailability throughout the crop season as per the crop demand is a potential challenge (Qin et 51 al. 2020). According to Fageria et al. (2011), the non-synchrony between crop N demand and 52 supply largely constrains rice production. In this context, sensor-based real-time N management 53 (RTNM) could be a potential approach to site-specific and demand-driven N management to 54 improve N use efficiency and reduce N losses (Huda et al. 2016). Several researchers have demonstrated strong positive associations between rice plant N status and leaf chlorophyll 55 56 content (Wood et al. 1993; Lin et al., 2010). Crop N nutrition as monitored by chlorophyll-meter 57 reading at different growth stages differs with cultivar, soil fertility, environmental conditions, 58 and locally adopted crop management practices (Mohanty et al. 2021). The precise adjustment of 59 chlorophyll-meter reading thresholds for fertilization scheduling with the optimal topdressing N 60 rate is therefore necessary in order to achieve maximum benefits of RTNM. Pre-sowing 61 application of manure in tropical soils can substantially contribute towards crop N nutrition, 62 depending on quantity, nutrient content, and rate of mineralization (Venkatesh et al. 2017). 63 Chlorophyll-meter based N scheduling would be a more realistic approach to N management in 64 integrated nutrient management module(s). However, the efficacy RTNM under integrated 65 nutrient management for yield advantage, nutrient use efficiency, and N economy module has 66 not been adequately studied. Further, variable N rates and timing as alter with different RTNM 67 treatments would provide valuable insights into N yield functions and relative weightage of N 68 application at different growth stages of rice crops.

69 In view of this, two years of calibration and one year of validation trials were conducted 70 to determine the best chlorophyll meter threshold for application scheduling and top-dressing N rate in lowland rice crop that is amended with farmyard manure at 5 t ha⁻¹ to maximize 71 72 productivity, nutrient use efficiency, and N economy and validate the best-fit RTNM treatment 73 across predominant rice cultivars. The major hypotheses of the study were (i). RTNM with 74 optimal combination(s) of N scheduling based on threshold chlorophyll-meter reading and top-75 dress N rate lead to yield advantage and higher nutrient use efficiency over FTNM in lowland flooded rice, (ii) temporal scheduling of fertilizer-N application (time of application), total 76 77 applied N rate, and frequency of N topdressing as differ within the RTNM treatments would 78 have a direct influence on rice productivity, (iii) the response scale of rice to SPAD chlorophyll-79 meter based RTNM would differ with cultivars having variable yield potential and crop duration, 80 and (iv) chlorophyll-meter based RTNM can be effectively integrated with manure amended 81 lowland rice production system for yield advantage and N economy.

82

83 Materials and Methods

84 Site characteristics

The experiments were conducted at the research farm of Bihar Agricultural University, Sabour, Bhagalpur ($25^{\circ}15'$ N; $87^{\circ}02'$ E, and 37 m a.s.l.). The site is located in the lower Indo–Gangetic plain (IGP) region (subtropical humid climate). The weather conditions during the crop seasons are presented in Supplementary Figure 1. The experimental soil is a silty–clay loam and classified as *Fluvisol* (IUSS Working Group WRB, 2015). The soil (0–0.15 m) had neutral pH 7.1, electrical conductivity of 0.47 dS m⁻¹, 4.9 g kg⁻¹ soil organic carbon, 12.2 mg kg⁻¹ of OlsenP and 66.9 mg kg⁻¹ of NH₄OAc–K (Jackson 1973). The soil available-N (98.2 mg kg⁻¹ KMnO₄–
N) of the field was lower than the critical soil N level (125 mg kg⁻¹ KMnO₄–N).

93

94 Experimental design and treatment details

95 Experiment 1 (Optimization trial)

96 In the optimization trial (rainy season 2013 and 2014), fourteen different N treatments were 97 evaluated in rice [cultivar (cv.) Rajendra Sweta (130-135 days duration)]. The treatments 98 comprised of one N control (no fertilizer-N application), one conventional practice of fixed-time 99 N management (FTNM) treatment, and twelve different SPAD 502 chlorophyll-meter (SCMR) 100 based real-time N management (RTNM) treatments using different combinations of SPAD 502 101 chlorophyll-meter reading (SCMR) based fertilization scheduling in which N was applied when 102 the SCMR of the youngest fully-extended leaf was less than or equaled 34 (SCMR₃₄), 36 103 (SCMR₃₆), 38 (SCMR₃₈), and 40 (SCMR₄₀), each with three top-dressing N rates (15, 20, and 25 104 kg N ha⁻¹). SCMR were recorded (n = 15) from each plot starting from the tiller initiation (20) 105 DAT) to the fully heading stage (90 DAT) at a 10-day interval using SPAD-502 chlorophyll-106 meter (Minolta Camera Co., Osaka, Japan). The FTNM treatment represents the current recommended dose of fertilizer-N (100 kg N ha⁻¹) application in four equal splits *i.e.* basal, active 107 108 tillering [40 days after transplanting (DAT)], panicle initiation (60 DAT), and heading (90 DAT) stages. The basal dose of fertilizer-N (i.e. 25 kg N ha-1) was applied to both the FTNM and 109 110 RTNM treatments and subsequent N was top-dressed as per treatments. The source of N was urea [CO(NH₂)₂]. Before the rice crop establishment, farmyard manure (FYM) at 5 t ha^{-1} (dry 111 112 weight basis) was applied to each plot. The quantity of N added with the FYM was 25.6 ± 0.7 kg ha⁻¹ and 27.7 \pm 1.1 kg ha⁻¹ (sample analyzed n = 3) in the year 2013 and 2014, respectively. The 113

114 treatments were allocated in a randomized complete block design (RCBD) with three 115 replications. The dimension of each plot was $6 \text{ m} \times 4 \text{ m}$. A filler space of 2.5 m was kept on all 116 sides of each plot.

117

118 Experiment 2 (validation trial)

In the validation trial (year 2015), the best-fit RTNM treatment along with a relatively lowyielding RTNM treatment (significant lower yield compared to the best-fit RTNM treatment) were identified/selected based on the results of the optimization trial (Experiment 1). These treatments were evaluated in seven high yielding rice cultivars *viz.* 27P31 (hybrid), RAU 724, MTU 7029, Rajendra Mahsuri, Rajendra Bhagawati, Rajendra Sweta and Arize 6444 (hybrid). Similar methods and management practices were employed as in the optimization trial.

125

126 Crop management

127 Rice seedlings were raised on a 20 m \times 5 m nursery bed (wet nursery). During the final field 128 preparation, well-decomposed FYM (5 t ha⁻¹) was added to each plot and thoroughly mixed. 129 Wet-tillage (puddling) was performed one day prior to transplanting. Twenty-two-day old 130 seedlings were transplanted on July 10 in 2013, July 15 in 2014 (optimization trial), and July 13 131 in 2015 (validation trial) at a spacing of 10 cm \times 15 cm. In both the experiments, 17.2 kg P ha⁻¹ 132 and 32.8 kg K were applied as a basal dose to all the treatments plots. From transplanting to 133 heading stage, continuous flooding/ponding conditions were maintained with irrigation. Weeds 134 were manually controlled by hand weeding 1-2 times. Necessary plant protection measures were 135 taken to prevent insect pests and diseases.

137 Yield and yield attributes

138 Yield attributing parameters like panicles m^{-2} , grains panicle⁻¹ and 1000–grain weight were 139 recorded from five rice hills randomly selected from each plot at the maturity stage. Grain and 140 straw yields were estimated from a net–plot area of 5.0 m × 2.0 m. The harvested crop was 141 threshed using a plot thresher and the produces was sun-dried. The total biomass of the crop was 142 estimated on an oven dry weight basis, while grain yield was adjusted at 12% moisture content 143 (w/w). Harvest index (HI) was calculated by the ratio of grain yield to total biological yield 144 (aboveground dry matter) and expressed as a percentage.

145

146 Plant nutrient analysis and calculation of N use efficiency

At maturity, plant and grain samples were collected, oven dried, ground and analyzed for total N content was determined by micro–Kjeldahl method (Yoshida et al. 1976). Agronomic N use efficiency (AE_N), N recovery efficiency (RE_N), internal N use efficiency (IE_N), partial factor productivity of applied N (PFP_N), nitrogen harvest index (NHI) and physiological N use efficiency (PE_N) were calculated by the following equations 1-6

152
$$AE_N (kg kg^{-1}) = \frac{GY_T - GY_C}{FN} \quad eq. 1$$

153
$$\operatorname{RE}_{N}(\operatorname{kg}\operatorname{kg}^{-1}) = \frac{\operatorname{NU}_{T} - \operatorname{NU}_{C}}{\operatorname{FN}} \quad \text{eq. 2}$$

154
$$IUE_N (kg kg^{-1}) = \frac{GY}{NU}$$
 eq. 3

155
$$PFP_N (kg kg^{-1}) = \frac{GY}{FN} \qquad eq. 4$$

156
$$\text{NHI (\%)} = \frac{\text{NU}_{\text{Grain}}}{\text{NU}_{\text{Total}}} \times 100 \quad \text{eq. 5}$$

157
$$PE_N (kg kg^{-1}) = \frac{AGB}{NU} \qquad eq. 6$$

where, $GY_T = grain$ yield in fertilizer-N treated plot (kg ha⁻¹), $GY_C = grain$ yield in N control plot (kg ha⁻¹), FN = quantity of fertilizer-N applied (kg N ha⁻¹), NU_T = total uptake of N in fertilizer-N treated plot (kg ha⁻¹), NU_C = total uptake of N in N control plot (kg ha⁻¹), GY = grain yield (kg ha⁻¹), NU = total uptake of N (kg ha⁻¹), NU_{Grain} = grain N uptake (kg ha⁻¹), NU_{Total} = Total uptake of N (kg ha⁻¹), AGB = total above ground biomass (kg ha⁻¹).

163

164 Calculation nitrous oxide emission

165 Total N₂O emission from each treatment plot was calculated indirectly following the 166 mathematical model (eq. 7) suggested by Tubiello et al. (2015) with the necessary modifications 167 at the regional scale and cropping period. A global warming potential equivalent factor 265 was 168 multiplied with the N₂O emission value to express the total emission as kg CO₂ equivalent ha⁻¹.

169 N₂0 emissions (kg ha⁻¹day⁻¹) = N × EF₁ ×
$$\frac{44}{28}$$
 eq. 7

where N₂O emissions from the fertilizer-N, manures, and residues added to the managed soil; N = N consumption from fertilizers and crop residues (kg N input year⁻¹); EF_1 = emitting factors of 0.01 for N₂O emissions from N input (kg N₂O–N kg⁻¹ N input).

173 Economic analysis

Costs and prices of all consumed inputs and outputs were considered in the economic analysis. All input cost components were summed up and denoted as working cost of cultivation, on which 3.5% interest, transportation and miscellaneous costs were added to calculate total variable cost of cultivation (TVCC). The total economic return from the grain and straw output was calculated as the gross return. The year-wise minimum support price of rice (paddy) (Government of India) and local market price of rice straw was used to convert the grain and straw yield to their economic value. Net return (NR) was calculated as the difference between gross return and TVCC. The ratio of net return to TVCC was denoted as the benefit-cost ratio
(BCR). Marginal return (MR) from N management treatments was computed using eq. 9.

183
$$MR (\$ ha^{-1}) = \frac{NR_T - NR_C}{TVCC_T - TVCC_C} \quad eq. 10$$

 $NR_T =$ net return with fertilizer-N treatment, $NR_C =$ net return from N control treatment, $TVCC_T$ = total variable cost of cultivation in fertilizer-N treatment, $TVCC_C =$ total variable cost of cultivation without fertilizer-N treatment (N control).

187

188 Statistical analysis

The data of optimization trial was analyzed following analysis of variance (ANOVA) of RCBD, while validation trial data were subjected to split–plot ANOVA (Cochran and Cox 1957). Principal component analysis was performed using statistical software PAST (version 3.26). The multivariate regression analysis was performed using 'Data Analysis Toolpak' Add–In of Microsoft Excel to explain the independent variables (X₁-Xn) as yield function and their corresponding coefficient weightage (C₁-Cn) [Y = C+ C₁X₁+ C₂X₂+.....+CnXn].

195

196 **Results**

Optimization trial (Experiment 1)

198 Fertilizer-N scheduling

The total fertilizer-N rate ranged between 55 kg (SCMR₃₄N₁₅) to 92 kg (SCMR₄₀N₂₅) N ha⁻¹ within the RTNM treatments (Table 1). The fertilizer-N rate was consistent over replications and cropping years, except for the treatment SCMR₄₀N₂₀ (Supplementary Table 1). The frequency of N splits ranged from 2 to 4 (SCMR₄₀N₁₅) (Table 1).

204 Crop response to N management treatments

205 The treatment SCMR₃₈N₂₀ had 8% (p < 0.05) higher yield than the FTNM treatment (Table 1). 206 The RTNM treatments SCMR₃₄N₁₅, SCMR₃₄N₂₀, SCMR₃₆N₁₅, and SCMR₃₆N₂₀ resulted in a 207 yield reduction (11-24%, p < 0.05) as compared to the FTNM, whereas the remaining RTNM 208 treatments had a similar yield to FTNM. The treatment SCMR₃₈N₂₀ had a lower panicle m⁻² but 209 equivalent grains panlcle⁻¹, stover yield and HI to the FTNM. Under fertilizer-N deficient 210 conditions (control), yield traits decreased (compared with FTNM) in the order of straw yield (-55%), grain yield (-45%), panicle m^{-2} (-43%), grains panicle⁻¹ (-17%) and 1000-grain weight (-211 212 10%), but increased HI (+11%) (p < 0.05). The highest HI was recorded in the treatment 213 SCMR₃₆N₂₅ (48.8%) followed by the N control treatment (47.1%). Grain yield showed higher 214 positive correlation with panicle density (r = 0.86) and straw yield (r = 0.80) (p < 0.05) (data not 215 presented). Under RTNM, a higher N rate increased yield in SCMR 34, 36, and 38, but not with SCMR 40. Similarly, increasing SCMR threshold increased yield under N rate (kg ha⁻¹) of 15 kg 216 217 (r = 0.996), 20 kg (r = 0.888), but not 25 kg N (r = 0.221) (data not presented).

218

219 N uptake and N use efficiency

The highest N uptake was recorded in the FTNM treatment (113 kg N ha⁻¹), while N uptake within the RTNM treatments ranged between 69 (SCMR₃₄N₁₅) to 112 (SCMR₃₈N₂₀) (Table 2). In the N control treatment, a greater magnitude of reduction (compared to FTNM treatment) was observed in the N uptake (62%) than in the scale of grain yield loss (45%). Strong positive correlations were observed between the N rate and N uptake in both the cropping years (r =0.911-0.938, p < 0.01) (Figure 1). Treatments SCMR₃₈N₂₀, SCMR₃₈N₂₅ led to the highest increase in AE_N over FTNM, while, except SCMR₃₈N₂₀, all the RTNM treatments had either equal or higher AE_N over FTNM (Table 2). The RE_N (kg kg⁻¹) was higher in treatments SCMR₃₈N₂₅ (0.87), SCMR₃₆N₂₅ (0.86), SCMR₃₈N₂₀ (0.84), which were significantly higher than the FTNM (0.70). The parameters IUE_N, PFP_N, NHI, and PE_N were higher in the RTNM treatments than the FTNM treatment. The highest IUE_N, NHI, and PE_N were recorded in the N control treatment (Table 2).

232 Multivariate analysis, economic return, and N₂O emission

Variable N rates, particularly at 50, 70, 80, and 90 DAT had a significant and higher positive influence on yield (p < 0.05). Likewise, among the variables N rate, N splits and NTI (an index that defines the earliness or delay in the application of fertilizer–N), only the N rate exhibited a significant positive influence on rice yield. There are marked year-to-year variations in the regression coefficient for the NTI (Figure 1). According to the PCA, treatment-induced vitiations in N splits, AE_N, RE_N, net return, and N rate. The strong positive association between total N rate and straw yield was evident from PCA results (Supplementary Figure 2).

The RTNM treatment SCMR₃₈N₂₀ led to the highest net return and benefit–cost ratio, which were +11% and +4% higher than the FTNM treatment (p < 0.01) (Table 3). With an increase in N top-dress rate, the marginal return was increased in SCMR₃₄ (up to 25 kg N ha⁻¹) SCMR₃₈ only (up to 20 kg N ha⁻¹). A reduction in seasonal N₂O emission (simulated) was estimated in all the RTNM treatments (2.4-12.9 kg CO₂ equivalent ha⁻¹) (p < 0.001) as compared to the FTNM treatment (N₂O GWP = 37.8 kg CO₂ equivalent ha⁻¹) (Figure 2)

246

247 **Experiment 2 (Validation trial)**

248 Treatment SCMR₃₈N₂₀ exhibited higher precision for N rate within replications over the 249 treatment SCMR₃₆N₂₀ (Supplementary Table 2). Notably, the yield difference between 250 treatments (SCMR₃₆N₂₀, SCMR₃₈N₂₀) for cv. Rajendra Sweta was 15.6% (p < 0.05) that was 251 almost similar to the optimization trial (13.9-28.7%, p < 0.05) (Table 4, Table 1). The yield 252 increment with the treatment SCMR₃₈N₂₀ over SCMR₃₆N₂₀ was recorded as > +10% for the *cvs*. 253 RAU724, Rajendra Sweta, and Rajendra Mahsuri, but remained comparable for cvs. MTU 7029, 254 Rajendra Bhagawati, and Arize 6444. The treatment $SCMR_{38}N_{20}$ had a higher straw yield and N 255 uptake than the treatment $SCMR_{36}N_{20}$ (Table 4, Table 5).

256

257 **Discussion**

258 **Experiment 1 (Optimization trial)**

259 The results show that SPAD-meter-based RTNM could be an efficient N management approach 260 to lowland flooded-rice production. Meanwhile, our results show that precise adjustment of 261 SCMR threshold for N scheduling along with optimal N top-dress rate are essential for yield advantage and higher N use efficiency over conventional FTNM. As, FYM (5 t ha⁻¹) was applied 262 263 to each treatment plot prior to crop establishment, therefore the organically bound N in manure 264 could have complemented the crop N requirement. In tropical climates, the faster mineralization 265 of organically bound N is readily accessible to the immediate crop; however, this may be 266 partially limited under flooded-rice soils due to anoxic soil conditions (Borase et al. 2020). The release kinetics of organically bound nitrogen can differ with crop stage, which is hard to 267 268 monitor. Therefore, RTNM could be a realistic approach to synchronize the demand and supply 269 of N based on relative chlorophyll meter readings and thereafter optimum scheduling of N 270 fertilization.

271 According to the results, not all RTNM treatments are effective in producing higher yield 272 but fertilizer savings is common in all treatments. Results indicate that RTNM following a lower 273 SCMR threshold (SCMR₃₄, SCMR₃₆) in combination with a low or medium top–dressed N rate 274 (15 and 20 kg N ha⁻¹) could lead to a yield loss (compared to the FTNM treatment), while, with a 275 higher top-dress N rate (25 kg N ha⁻¹) had attained a similar yield to conventional FTNM 276 treatment with a saving of 25 kg N ha⁻¹ (Table 1). This result suggests that a higher N 277 topdressing rate is essential to have an extended effect of top-dress N in RTNM following a SCMR threshold below 38. A topdressing N rate of 20 kg N ha⁻¹ in combination with fertilizer-N 278 279 scheduling with SCMR₃₈ could produce higher yields, fertilizer-N use efficiency, and higher 280 economic return compared to FTNM (Tables 1, 2, and 3), and was therefore recommended as a 281 best-fit RTNM treatment for the medium duration rice cv. Rajendra Sweta. The positive impact 282 of the treatment $SCMR_{38}N_{20}$ was possibly attributed to synchronous and balance N application as 283 per crop demand. The N fertilization schedules in the $SCMR_{38}N_{20}$ treatment are basal, 50, 70, 90 284 DAT, while in the FTNM treatment they are basal, 40, 60, 90 DAT (Supplementary Table 1, 285 Supplementary Figure 3), and thus, the delayed application of topdressing fertilization may have 286 been the cause for balanced growth, whereas the higher N rates applied in the FTNM treatment 287 resulted in higher vegetative growth, but no apparent increase in yield attributing traits. The 288 multivariate regression analysis results indicate that the variable N application at 50, 70, 80, and 289 90 DAT had a direct positive influence on yield, which closely matched with the best-fit RTNM 290 treatment SCMR₃₈ N_{20} (Figure 1). The consistency in the results within the cropping year also 291 validates the regression model. The treatment SCMR₃₈N₂₀ had a lower panicle density than the FTNM treatment, whereas the number of grains panicle⁻¹ and seed weight was higher in the 292 293 treatment SCMR₃₈ N_{20} over the FTNM treatment (Table 1), and this finding specifically hints a

higher intra-plant competition in the FTNM treatment possibly attributed to increased tiller production that might have reduced number of grains panicle⁻¹ and their weights. The reduction in yield determining traits like grains panicle⁻¹ and seed weight are mostly genetically configured, but an imbalance in growth (higher biomass growth, (Supplementary Figure 4) may lead to affect source-sink balancing, which is further confirmed with the results of HI. Wei et al. (2018) also reasoned that the source-sink relationship during grain filling is not merely a matter of carbon supply versus demand, but highly coupled with grain N demand.

Results indicate that the RTNM following a higher threshold SCMR (> 38) could increase topdressing frequency but not necessarily increase yield over the FTNM. This is especially true when N splits are not uniformly distributed (temporal) during the crop season. Additionally, faster growth may result in the dilution of chlorophyll in leaf tissue, misleading the SCMR-based RTNM.

306 Besides, a faster growth may cause dilution of chlorophyll in leaf tissue and thereby 307 mislead the SCMR-based RTNM. Concerning the increased yield benefits from N application at 308 the later growth stages (50-90 DAT) a dynamic approach of N management (lower N topdressing 309 rate at the initial growth stages and higher rate at the later growth stages) would result in further 310 up-scaling of N use efficiency. Peng et al. (2006) also suggested that a quantum leap in AE_N is 311 possible in the intensive rice-growing areas in China by simply reducing the current N rate and 312 by allocating less N at the early vegetative stage. PCA reveals that the frequency of N split 313 application is not significantly correlated with higher N use efficiency (especially AE_N) as a 314 right-angle relationship between variables (or traits) represents a non-significant correlation 315 (Supplementary Figure 2). A trait vector length represents the sensitivity scale of a variable to N 316 management treatments, and the results suggest that the benefits of RTNM are much more

317 prominent on N use efficiency parameters (AE_N and RE_N) than yield and its attributes. The 318 prominent year-wise deviations in the regression coefficient of NTI could be attributed to the 319 seasonal variation in weather conditions (particularly rainfall) and mineralization rate (Figure 1).

320 The RTNM treatment SCMR₃₈N₂₀ exhibited higher AE_N and RE_N over the FTNM 321 treatment, which is due to the yield gain over the FTNM treatment as well as reduced use of 322 fertilizer-N. This study corroborates the findings of Khurana et al. (2008) and Singh et al. (2002) 323 in western parts of India with a SPAD critical value of 37.5. The increased IUE_N and PE_N with 324 higher yield certainly indicate improved intra-plant nutrient use efficiency and N remobilization 325 from source to sink in the treatment $SCMR_{38}N_{20}$ over the FTNM. The study also indicates the 326 advantage of RTNM in reducing the NO₂ emission from a lowland flooded rice soil primarily by 327 reducing the fertilizer-N input. Lowland flooded rice ecosystems contribute largely toward total 328 agricultural greenhouse gas emissions. Fertilizer-N applied to flooded rice crops is much prone 329 to several losses such as N leaching, denitrification etc., which ultimately downscale the N use 330 efficiency. The imbalance in N fertilization scheduling and a higher rate of N topdressing might 331 have resulted in higher N losses for FTNM treatment, whereas higher N application at early 332 stages might have led to higher N losses.

A higher topdressing frequency with a low top-dress N rate is considered an ideal approach to upscale yield and N use efficiency; however, the economic feasibility must be taken into consideration to determine the optimum N splits. For instance, chlorophyll-meter-based RTNM following a higher SCMR as a threshold could increase the frequency of N topdressing particularly when the N topdressing rate is low as observed for the treatment SCMR₄₀N₁₅ (Table 1). The treatment SCMR₄₀N₁₅ recorded a comparable grain yield (5204 kg ha⁻¹) to the treatment 339 SCMR₃₈N₂₀, however, the net return was higher in the SCMR₃₈N₂₀ treatment over the treatment 340 SCMR₄₀N₁₅ (Table 5) and therefore SCMR₃₈N₂₀ was selected as the best treatment.

The savings of fertilizer-N and a higher yield in best-fit RTNM treatment (SCMR₃₈N₂₀) led to favourable production economics. The researchers noted the exorbitant cost of rice cultivation was due to the overuse of fertilizer–N (Koch et al. 2004). The economics were calculated based on the government-subsidised price of urea (Government of India). Therefore, in reality, the cost involved in N management would be higher than the estimated cost of fertilizer N in the study. Thus, the margin of economic return between the best-fit RTNM treatment and FTNM with unsubsidised fertilizer-N (urea) price would be more significant.

348

349 **Experiment 2 (validation trial)**

350 The results demonstrate a need for cultivar-specific calibration of optimal SCMR for fertilization 351 scheduling in lowland flooded rice. According to the validation trial result, the test cv. Rajendra 352 Sweta showed a similar response to that of the optimization trial, and thus, the result verifies the 353 field scale efficacy/precision of RTNM treatments for a particular rice cultivar in lowland 354 ecologies. Nevertheless, the yield advantage with best-fit RTNM treatment varied greatly across 355 cultivars. Differences in yield potential, N demand, and crop duration might explain the 356 differential response of cultivars to RTNM. Given the differential response of rice cultivars to 357 the selected RTNM treatments, a cultivar-specific RTNM recommendation would be essential to 358 have the best results. Additionally, the results indicate that the precision of RTNM treatment for 359 the applied N rate and timing within replicates may be equally considered for precise adjustment 360 and validation of cultivar and location-specific N management in rice. Shukla et al. (2004) also

reasoned that site-specific nutrient management must consider the wider variability in leaf colourtraits within the cultivars.

363

364 **Conclusions**

365 The study concluded that SCMR-based RTNM could be a sustainable approach of N 366 management in lowland flooded-rice systems over the conventional FTNM. However, precise 367 adjustment of the SCMR threshold in combination with optimal N rate is vital to realize the 368 optimal benefits of RTNM. The results suggested that SCMR based RTNM in lowland flooded 369 rice could be effectively applicable under integrated nutrient management module(s). The optimization trail results revealed that basal application of 25 kg N ha⁻¹ followed by top-dressing 370 of fertilizer-N at 20 kg N ha⁻¹ when the SCMR \leq 38 could improve yield, N use efficiency, net 371 372 return and reduced fertilizer–N rate over FTNM. The yield gain in the treatment $SCMR_{38}N_{20}$ 373 over the FTNM was primarily attributed to improved source-sink balance with a timely (delayed) 374 supply of N during crop season. The regression analysis results explained that N application at 375 later growth stages (panicle initiation to full heading stage) had a higher scale of impact on rice 376 yield over the early applications. Therefore, a variable N top-dress rate [a lower dose at early 377 growth stage(s) and a higher in the later growth stages] would further improve the N economy in 378 RTNM in lowland rice, which is to be ascertained with systematic studies. The study suggests 379 that there is a need for cultivar-specific calibration of SCMR threshold limit in combination with 380 optimal top-dress N rate for maximum benefits.

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 Table 1. Effect of different N management practices on yield attributes of rice (pooled data of

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2013 and 2014) [Experiment 1].

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Treatment	Fertilizer-N applied (kg ha ⁻¹)	N split (nos.)	Panicles m ⁻² (nos.)	Grains panlcle ⁻¹ (nos.)	Test weight (g)	Grain yield (kg ha ⁻¹)	Straw yield (kg ha ⁻	Harv index (%)
SCMR ₃₄ N ₁₅	55	2	235 ^e	73.3 ^{cd}	22.9 ^{cd}	3753 ^e		45.9 ^b
SCMR34N20	65	2	260 ^d	83.1 ^a	23.7 ^{abc}	4164 ^d	4992 ^{fg}	45.5 ^c
SCMR ₃₄ N ₂₅	75	2	281 ^{bcd}	79.7 ^{abc}	24.1 ^{ab}	4893 ^{bc}	5365 ^{ef}	47.7 ^a
SCMR ₃₆ N ₁₅	63	3	256 ^d	74.5 ^{bcd}	22.9 ^{cd}	4132 ^d	4867 ^{gh}	46.1 ^b
SCMR ₃₆ N ₂₀	65	2	263 ^d	78.7^{abc}	24.1 ^{ab}	4412 ^c	4983 ^{fg}	46.9ª
SCMR ₃₆ N ₂₅	71	2	271 ^{cd}	80.8 ^{ab}	24.5 ^a	4767 ^c	5008 ^{fg}	48.8 ^a
SCMR38N15	70	3	280 ^{bcd}	75.8 ^{bc}	23.3 ^{bc}	4764 ^c	5496 ^{def}	46.4 ^b
SCMR ₃₈ N ₂₀	82	3	280^{bcd}	84.9 ^a	24.6^{a}	5323 ^a	6277 ^{bc}	46.0^{b}
SCMR ₃₈ N ₂₅	75	2	284 ^{bc}	75.9 ^{bc}	24.6 ^a	5106 ^{abc}	5745 ^{cde}	46.9 ^a
SCMR40N15	85	4	282 ^{bc}	82.8 ^{ab}	24.5 ^a	5204 ^{ab}	5927 ^{bcd}	46.8ª
SCMR40N20	85	3	286^{abc}	79.3 ^{abc}	23.7^{abc}	5005 ^{abc}	6545 ^a	43.2 ^e
SCMR40N25	92	3	290 ^{ab}	79.6 ^{abc}	24.0^{abc}	4862 ^c	6340 ^{ab}	43.6 ^d
FTNM	100	3	303 ^a	81.2 ^{ab}	24.2^{ab}	4952 ^{bc}	6800 ^a	42.3 ^e
Control	-	-	173 ^f	67.5 ^d	21.7 ^d	2700 ^f	3045 ⁱ	47.1ª
LSD (<i>p</i> = 0.05)	7.4	-	17.5	7.2	1.2	346	563	2.4
Pooled ANOVA	(<i>p</i> value)		0.047	0.050	0.1.15	0.010	0.056	0.004
Year	1.00		0.065	0.052	0.147	0.013	0.056	0.004
Treatment	< 0.001		< 0.001	0.002	< 0.001	< 0.001	< 0.001	< 0.0
Year × Treatment	0.124		0.150	0.564	0.309	0.002	0.027	< 0.0
<i>SCMR</i> , SPAE <i>FTNM</i> , Fixed are significant	502 chloroph time N manag	hyll met gement. $a p \le 0.05$.	er reading <i>1-i</i> , treatme	; <i>N</i> , fertilize ents with dif	er-N topdre ferent lowe	essing rate ercase supe	(kg N ha rscript lett	⁻¹); ers
Table 2. Effe nitrogen recov applied N (PF (pooled data c	ect of differen very efficiency P _N), nitrogen h of years 2013 a	t N man (RE _N), i narvest in nd 2014)	agement p nternal N idex (NHI) [Experime	practices on use efficienc and physiol ent 1].	agronomic y (IE _N), pa ogical N us	N use efficiences of the second secon	ficiency (A productiv y (PE _N) o	AE _N), ity of f rice

	uptake	(kg kg^{-1})	(kg kg^{-1})	(kg kg^{-1})	(kg kg^{-1})	(%)	(kg kg^{-1})
	(kg ha ¹)						
SCMR ₃₈ N ₂₀	69.0 ^f	19.1 ^d	0.47 ^h	54.3 ^b	68.2 ^a	0.64^{ab}	125 ^b
SCMR ₃₈ N ₂₅	80.7 ^e	22.5 ^{cd}	0.57 ^g	51.6 ^{bc}	64.1 ^{abc}	0.62 ^{bc}	120 ^{bc}
SCMR34N25	103.7 ^{bc}	29.2 ^{ab}	0.80^{bc}	47.2 ^{efg}	65.2 ^{ab}	0.59 ^{de}	115 ^{cdef}
SCMR ₃₆ N ₁₅	81.4 ^e	22.5 ^{cd}	0.61 ^g	50.7 ^{cd}	66.6 ^a	0.62 ^{bc}	121 ^{bc}
SCMR ₃₆ N ₂₀	89.9 ^d	26.3 ^{bc}	0.72 ^{ed}	49.0 ^{cde}	67.9 ^a	0.61 ^{cd}	115 ^{cdef}
SCMR ₃₆ N ₂₅	104.0 ^{bc}	28.9^{ab}	0.86^{ab}	45.8 ^{fgh}	67.6 ^a	0.58 ^{ef}	112 ^{def}
SCMR ₃₈ N ₁₅	97.6 ^c	29.5 ^{ab}	0.78 ^{cde}	48.8 ^{cdefg}	68.1 ^a	0.61 ^{cd}	119 ^{bc}
SCMR38N20	111.6 ^a	32.2 ^a	0.84^{abc}	47.8 ^{defg}	65.6 ^{ab}	0.60 ^{cde}	112 ^{def}
SCMR ₃₈ N ₂₅	108.7^{ab}	32.1 ^a	0.87 ^a	46.8 ^{efg}	68.1 ^a	0.60 ^{cde}	$110^{\rm f}$
SCMR ₄₀ N ₁₅	110.0 ^{ab}	29.4 ^{ab}	0.78^{cde}	47.4 ^{efg}	61.2 ^{bc}	0.59 ^{de}	118 ^{cd}
SCMR40N20	107.0^{ab}	26.7 ^{bc}	0.74^{def}	47.0 ^{efg}	59.2 ^{cd}	0.59 ^{de}	119 ^{bc}
SCMR40N25	107.5 ^{ab}	23.6 ^c	0.70^{f}	45.5 ^{gh}	54.0 ^{de}	0.56^{fg}	117 ^{cde}
FTNM	113.3 ^a	22.5 ^{cd}	0.70^{f}	43.7 ^h	49.5 ^e	0.55 ^g	111 ^{ef}
Control	43.4 ^g	-	-	62.2 ^a	-	0.67 ^a	142 ^a
LSD (<i>p</i> = 0.05)	7.08	4.4	0.07	3.1	5.3	0.03	6.8
Pooled ANOVA	(p value)						
Year	1.00	0.019	0.034	0.054	0.006	0.119	< 0.001
Treatment	< 0.001	< 0.001	< 0.001	< 0.001	< 0.001	< 0.001	< 0.001
Year × Treatment	0.124	0.039	0.003	0.309	0.563	0.659	0.320

SCMR, SPAD 502 chlorophyll-meter reading; *N*, topdressing N rate (kg N ha⁻¹); *FTNM*, fixed-495 time N management; *DAT*, days after transplanting. *a-h*, treatments with different lowercase 496 superscript letters are significantly different at $p \le 0.05$.

502 Table 3. Effect of different N management practices on economics of rice production (pooled503 data of years 2013 and 2014) [Experiment 1].

Treatment	Total variable cost of cultivation (\$ ha ⁻¹)	Gross return (\$ ha ⁻¹)	Net return (\$ ha ⁻¹)	Benefit cost ratio	Marginal return (\$ ⁻¹ ha ⁻¹)
SCMR ₃₄ N ₁₅	454 ^d	697 ^f	243 ^f	1.53 ^e	2.19 ^d
SCMR ₃₄ N ₂₀	465 ^c	774 ^e	309 ^e	1.66 ^d	2.77 ^c
SCMR34N25	483 ^b	903 ^{bcd}	420 ^{cd}	1.87 ^{bc}	3.44 ^{ab}
SCMR ₃₆ N ₁₅	465 ^c	767 ^e	302 ^e	1.65 ^d	2.67 ^c
SCMR ₃₆ N ₂₀	470 ^c	816 ^e	346 ^e	1.73 ^d	3.06 ^{bc}
SCMR ₃₆ N ₂₅	480 ^b	876 ^{de}	397 ^d	1.82 ^c	3.30 ^{bc}

SCMR ₃₈ N ₁₅	481 ^b	883 ^{bcd}	402 ^d	1.83 ^{bc}	3.31 ^{bc}
SCMR38N20	495 ^a	989 ^a	494 ^a	2.00^{a}	3.76 ^a
SCMR ₃₈ N ₂₅	487 ^b	944 ^{abc}	457 ^{abc}	1.93 ^{ab}	3.68 ^{ab}
SCMR ₄₀ N ₁₅	496 ^a	964 ^{ab}	468 ^{ab}	1.94 ^{ab}	3.47 ^{ab}
SCMR ₄₀ N ₂₀	484 ^b	938 ^{abcd}	455 ^{abc}	1.93 ^{ab}	3.78 ^a
SCMR ₄₀ N ₂₅	482 ^b	911 ^{bcd}	430 ^{bcd}	1.89 ^{bc}	3.59 ^{ab}
FTNM	486 ^b	933 ^{abcd}	447 ^{bc}	1.92 ^b	3.62 ^{ab}
Control	392 ^e	499 ^g	107 ^g	$1.27^{\rm f}$	-
LSD ($p = 0.05$)	7.7	62.2	44.8	0.08	0.45
Pooled	(<i>p</i> value)				
ANOVA					
Year	0.015	0.018	0.018	0.020	0.147
Treatment	< 0.001	< 0.001	< 0.001	< 0.001	< 0.001
Year ×	< 0.001	0.003	0.005	0.013	0.029
Treatment		0.005	0.005	0.015	

SCMR, SPAD 502 chlorophyll-meter reading; *N*, topdressing N rate (kg N ha⁻¹); *FTNM*, fixed-507 time N management. *a-g*, treatments with different lowercase superscript letters are 508 significantly different at $p \le 0.05$.

Table 4. Crop yields, N uptake, and partial factor productivity of applied N (PFP_N) of different

516 rice cultivars to RTNM treatments [Experiment 2].

Cultivar	Crop duration	Grain yield (kg ha ⁻¹)		Straw yield (
	(days)	$RTNM_1$	RTNM ₂	Mean	$RTNM_1$
27P31	128-132	5920 ^{B#}	6467 ^A	6194 ^d	13854 ^B
RAU724	~150	5630 ^B	6480 ^A	6055 ^d	13595 ^B
MTU7029	~150	6487 ^A	6770 ^A	6629 ^c	14297 ^B
Rajendra Sweta	130-135	4127 ^B	4753 ^A	4440 ^e	10418 ^A
Rajendra Bhagawati	120-125	4416 ^A	4260 ^A	4338 ^e	10246 ^A
Rajendra Mahsuri	~150	6612 ^B	7330 ^A	6971 ^a	14647 ^B
Arize 6444	135-140	6727 ^A	6683 ^A	6705 ^b	14231 ^A
Mean		5702 ^B	6106 ^A		12755 ^B
LSD ($p = 0.05$)					
Cultivar (C)	-	194			319
N treatment (N)	-	351			796
$C \times N$ interaction (<i>p</i> value)	-	< 0.001			0.001

519 A-B, different uppercase superscript letters indicates significant difference within the RTNM

520 treatments at $p \le 0.05$. RTNM₁ = SCMR₃₆N₂₀, RTNM₂ = SCMR₃₈N₂₀. *a-e*, treatments (cultivars)

- with different lowercase superscript letters indicates significant difference at $p \le 0.05$ based on
- the LSD value of split plot ANOVA. # two-sample *t*-test at $p \le 0.05$ was employed to compare the significant difference within the N management treatments for individual cultivar.

Table 5. Crop yields, N uptake, and partial factor productivity of applied N (PFP_N) of different

528 rice cultivars to RTNM treatments [Experiment 2].

Cultivar	Fertilizer	Fertilizer N applied (kg ha ⁻¹)			N uptake (kg ha ⁻¹)		
	$RTNM_1$	RTNM ₂	Mean	$RTNM_1$	RTNM ₂	Mean	RTN
27P31	78.3 ^A	85.0 ^A	81.7 ^b	117.5 ^A	124.8 ^A	121.2 ^c	75.6
RAU724	71.7 ^A	85.0 ^A	78.4 ^b	98.1 ^B	128.7 ^A	113.4 ^d	78.5
MTU7029	78.3 ^A	85.0 ^A	81.7 ^b	128.2 ^A	131.2 ^A	129.7 ^b	82.8
Rajendra Sweta	71.7 ^A	85.0 ^A	78.4 ^b	89.3 ^A	95.3 ^A	92.3 ^e	57.6
Rajendra Bhagawati	78.3 ^A	85.0 ^A	81.7 ^b	88.9 ^A	86.3 ^A	87.6 ^e	56.4
Rajendra Mahsuri	105.0 ^A	91.7 ^A	98.4 ^a	127.0 ^B	142.6 ^A	139.8 ^a	63.0
Arize 6444	85.0 ^A	78.3 ^A	81.7 ^b	128.2 ^A	123.0 ^A	125.6 ^{bc}	79.1
Mean	81.2 ^A	85.0 ^A		111.0 ^B	118.9 ^A		70.2
LSD ($p = 0.05$)							
Cultivar (C)	9.9			6.08			9.1
N treatment (N)	ns			11.4			ns
$C \times N$ interaction (<i>p</i> value)	ns			ns			ns

A-B, different uppercase superscript letters indicates significant difference within the RTNM treatments at $p \le 0.05$. RTNM₁ = SCMR₃₆N₂₀, RTNM₂ = SCMR₃₈N₂₀. *a-e*, different lowercase superscript letters indicates significant difference within the cultivars at $p \le 0.05$ based on the LSD value of split plot ANOVA. # two-sample *t*-test at $p \le 0.05$ was employed to compare the significant difference within the N management treatments for individual cultivar.



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Figure 1. Weightage of regression coefficient of N scheduling at different growth stages (independent variabales) to explein their relative influence on rice grain yield (dependent variable) based on linear mixed regression model (**a**, **b**). Weightage of regression coefficient of N rate, N splits and NTI variables (independent variabales) to explain their relative influence on rice yield (**c**, **d**). Linear relationship between fertlizer-N input and N uptake in rice in the year 2013 and 2014 (Experiment 1, Optimization trial) (**e**, **f**). NTI, N timing index defines the earliness or delay in the application of fertilizer-N. DAT, days after transplanting. * p < 0.05; ** p < 0.01; *** p < 0.001.



Figure 2. Simulated estimates of the reduction in N_2O emission (projected) in different RTNM treatments compared to the FTNM treatment (Experiment 1).