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2 **Relative abundance of insect predators varies among rice plots as a function of**
3 **surrounding landscape**
4

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16

17 **Abstract**

18 Relationships among the population abundance of four predator groups for rice insect pests,
19 namely: carabid beetles, staphylinid beetles, green mirid bugs, and spiders in three landscape
20 categories were evaluated. Both rice plots and the associated bund margins of these rice plots
21 found among three Bangladesh landscape categories were sampled by sweep net. The results
22 revealed that the abundance significantly varied across landscapes. The rice landscape of one
23 location harbored higher numbers of a specific predator than other location in other regions of
24 Bangladesh. The results also showed a dependency on the width of the rice bund margins of
25 the rice plots, where spiders populations increased with increased bund widths, but the
26 population abundance of these predators did not depend on the diversity of the number of
27 weed species found on the rice bund margins. The relative abundance of predator populations
28 also significantly differed among the three landscapes, with the green mirid bug having the
29 highest number among the four predators. This study indicates that predators of rice insect
30 pests are highly landscape specific. In order to design integrated pest management systems
31 for different Bangladeshi rice production locales, considerations unique to the characteristics
32 of each locale are necessary. Preliminary efforts to apply variography analyses to the RED
33 spectral band of LANDSAT 8 imagery from December 2016 are presented as first step

34 toward learning a suite of methods which describe useful local characteristics affecting rice
35 pest predators.

36 **Keywords:** Rice landscape, natural enemies, location, population dynamics, variography,
37 LANDSAT 8

38

39 **Introduction**

40 Rice (*Oryza sativa* L.), the staple diet of over half of the world's population, is grown
41 on over 158 million hectares worldwide, which produced over 465 million tons in 2012. In
42 Bangladesh, rice occupies about 77% of the cropped areas (Bhuiyan *et al.* 2004) which
43 account for a total 11.6 million hectares to produce 34 million tons of milled rice (IRRI,
44 2014). Bangladesh has 3 rice growing seasons (BRRI, 2016); namely, *Aus* (monsoon rice),
45 *Aman* (rain-fed with supplemental irrigation) that has two types of production (Broadcasted
46 Aman and Transplanted Aman), and the *Boro* (irrigated and well managed rice) season. Rice
47 is cultivated throughout the year and the intensity of cultivation increases day by day to meet
48 demands from more people living in Bangladesh every year. The rice agro-ecosystem covers
49 the major part of the non-urban land area in Bangladesh. These rice eco-systems are inhabited
50 by hundreds of arthropod species (Heong 2011) which perform various ecological functions,
51 such as herbivory (feeding on the rice plants), predation, parasitization, pollination,
52 decomposition, and nutrient cycling.

53 To date, 232 rice insect pests and 375 beneficial arthropod species have been
54 identified from the rice ecosystem in Bangladesh (Ali *et al.*, 2017). However, while fewer
55 than 20 species are of significance in causing yield losses when they occur in sufficiently
56 large numbers in India, Bangladesh, numbers up to 20-33 total species considered of
57 importance to cause economic damage to rice production (BRRI, 2007). These pests, in turn,
58 are subjected to attack by predators and parasitoids, and thus, are often naturally kept in

59 check. This intricate food web of relationships among the rice plants, the pests, and the rich
60 biodiversity of natural enemies constantly strives to maintain an equilibrium preventing
61 abnormal increases in abundance of pest species. This equilibrium level of a rice ecosystem is
62 also often broken due to heavy use of synthetic fertilizers and pesticides. The breakdown in
63 'ecological resilience' of a rice farm induces pest's outbreaks (IRRI, 2011) that causes
64 economic damage to rice growers worldwide, despite usage of insecticides.

65 Scientists have long claimed that indiscriminate use of pesticide is the main reason for
66 major outbreaks of insect pests in many kinds of crop production plots (Stern et al., 1959). A
67 recent example in rice is the increase in outbreak frequency of *Nilaparvata lugens* (brown
68 planthopper) over numerous Asian rice growing countries in recent years, 2005–2012
69 (Heong, 2010), due to adverse effects on planthopper natural enemies (NE) caused by
70 increased use of broad spectrum insecticides for control of other kinds of rice pests (Chien
71 and Cuong, 2009, Heong, 2009, Islam and Haque, 2009, Moni, 2011;2012, Luecha, 2010,
72 Matsumura and Sanada-Morimura, 2010, Soitong and Sriratanasak, 2012, Teo, 2011). The
73 use of pesticides increased in Bangladesh by 200% from 1997-2000, 250% by 2006 and by
74 nearly 500% by 2014 (BPCA, 2015). More than half of the amount of those insecticides was
75 applied against rice pests. Synthetic chemical-like insecticides are hazardous and harmful for
76 non-target organisms (e.g., Travisi et al., 2006, Ahmed et al., 2002; 2011). Besides just the
77 increasing usage of chemical insecticides over the years, other factors such as climate change
78 and landscape change also induce disappearance of NE from Bangladesh rice plots. Analysis
79 of

80 Natural enemies abundance in a crop plot vary with the apportionate areas of different
81 habitats that prevail in the surrounding landscape (Bianchi et al., 2006; Werling et al., 2011)
82 and possibly with differences in small scale changes in timing and choice of management
83 decisions (Willers et al. 2005; Shaw and Willers, 2006) for another insect pest, such as *Lygus*

84 *lineolaris* in Mid-South, USA, cotton. Landscape changes, which shape the habitat structure,
85 materials, and biotic interactions in agroecosystems, are also important factors driving pest
86 and NE population abundance (Woltz et al. 2012). Landscape characteristics can influence
87 pest and NE population in crop field. Recently, remote sensing method provide quickly
88 various types of surface monitoring data such as local, landscape, vegetation, specific crop,
89 water body and animal population. Variogram analysis of the data for spatial heterogeneity
90 can help identify explain related ecological phenomena (Zhaofei et al. 2013). Here we intend
91 to use variogram analysis and find their impact on landscape characteristic which also can
92 explain the pest and NE population abundance. It could also be used to improve quantitative
93 agricultural remote sensing monitoring in a spatial heterogeneity area as well.

94 Examining such kinds of impacts on pests or NE and understanding their mechanisms
95 may help to design pest management strategies at landscape scales (Wang et al. 2015).
96 However, the description of the abundance of NE in different Bangladesh rice landscape
97 categories remains elusive. Therefore, the objective of this study was to assess the abundance
98 of NE in different rice landscapes to help design pest management strategies influenced by
99 the different rice production seasons, types of production styles (i.e., small, household
100 farmers vs large, non-household farmers) and categories of landscape-scale agro-ecosystems.
101 In addition, variograms are used to analyze the landscape characterize based on LANDSAT 8
102 images collection. To target this objective we surveyed, recorded, and summarized the
103 abundance of several NE species from different rice landscapes in Southern Bangladesh.

104 **Materials and Methods**

105 The sampling experiments were conducted in three categories of landscape located in
106 Southern Bangladesh, or the Barisal Division, which includes within it the Barisal and
107 Jhalkathi Districts, where each district is further divided into two more smaller governmental
108 entities, known as the Upazila, and within these, the smallest one, known as a Union (Refer to

109 Fig. 1, where all but the Upazila are mapped). Each landscape category represents a unique
110 type within several Unions of these two Districts, and which are sufficiently describable to
111 permit replication (in geographic space at the Union designation) of each category. We
112 defined these rice landscapes within a buffer zone around each one, into three category
113 classes based on surrounding features and characteristics found within the buffer zone
114 surrounding candidate rice plots. Each landscape was replicated two times.

115 Landscape I — Rice plots in this category are typically surrounded by big and small
116 fruit trees, or forests (such as deciduous and coniferous trees). The entire rice plots, here, are
117 enclosed by densely perennial habitats, having fewer kinds of annual crops and lower
118 vegetational diversity. The perennial habitats are found in close proximity to rice plots, with
119 an range of distances from 10-30 m. The main feature of this landscape is the presence of
120 small, narrow (so called canal), drainage flows between the rice plots and (concrete) roads,
121 which always flows, and with some weeds growing in the canals. The canals are connected to
122 rice plots, which were very muddy types. The irrigation system of this landscape is comprised
123 of a shallow tube well. The width of rice bunds surrounding rice plot, are between 25-35 cm
124 and separate one smallholder plot from another one. Landscape I replicates were selected
125 near Protap, in the Rajapur Upazila Union, and the Jhalokathi Sadar Union of southern
126 Bangladesh, and were of similar structure and composition by visual interpretations. The
127 specific location of this landscape is presented in Fig. 1A.

128 Landscape II — This category consists of rice plots, homestead trees, and a rain forest,
129 containing both small and big trees. The landscapes are located the Nalchity Upazila, in the
130 Jhalkati and Barisal Sadar Unions of southern Bangladesh, characterized by less muddy rice
131 plots found near roadsides, with a few fruit and forest trees found around the rice plots and
132 planted along the roadside. Any nearby perennial habitats were very far from the rice plots,
133 compared to Landscape I, with distances between 100-200 m. There are no canal or drainage

systems in this landscape and the irrigation system is comprised of deep tube wells. The specific location of these landscape replicates is also shown in Fig. 1A.

Rice plots from each landscape selected for study ranged 15-20 ha in area. The total set of rice plots in each replicate was composed of 50-70 plots separated from each other by bunds in each sample location among the landscape categories, except for Landscape III. In the Landscapes I and II, each small plot was occupied by one smallholder rice farmer, whom maintained their plots according to improved rice production technology. The size of farmers plot ranged 1500-2500 m². The plot size in Landscape III ranged between 1000 - 2000 m², and were farmed by smallholder farmers. The 'plots' for all replicates selected to record arthropod populations was transplanted with BRRI dhan29 (a mega-rice variety in Bangladesh). Four-6 plots from each landscape were considered for data collection.

156 For transplanting BRRI dhan29 in selected rice plots, farmers raised seedlings in a
157 seedbed. Seedbed management was performed according to the traditional farm practices
158 (BRRI, 2011). Before transplanting seedlings into candidate sample plots, land was well

159 prepared according to the common practice of wetland soil preparation followed by
160 laddering. Laddering is the cultural farming practices where the ladder is used to break down
161 clods and level the plot once or twice after ploughing. Thirty-five to 45 d old seedlings were
162 transplanted in selected rice plots during the *Boro* rice cultivation season in 2015-2016.
163 Standard transplanting space (20 x 20 cm²) was maintained. Fertilizers containing N, P, K,
164 and S were applied at the rates of 82, 15, 38, 10.6 and 2.7 kg ha⁻¹ respectively, using urea,
165 triple superphosphate (TSP), muriate of potash (MOP), and gypsum. The total amount of
166 TSP, MOP, gypsum, and 1/3 of the urea amount were applied during the final land
167 preparation period. The remaining urea was top dressed in two equal splits at 20 d after
168 transplanting (DAT) (or the early tillering stage) and 40 DAT (or the maximum tillering
169 stage), synchronized with irrigation or wet soil conditions, because the sampling experiment
170 was conducted under irrigation conditions. Pesticides were applied two times in all rice plots
171 across the tested landscapes. Therefore similar amount of pesticides were received by each
172 rice plot.

173 Arthropod populations were recorded from 4-6 plots of each replicated rice landscape
174 category which means 24-36 plots (4-6 plots x 6 landscapes). Two landscapes of each
175 category were considered for this study. The sampling was conducted two times. A sampled
176 plot was randomly selected from the entire set of rice plots found, as present in a replicated
177 landscape category. Arthropod populations were collected from both the chosen rice plot, and
178 its adjoining rice bund, using a sweep net. Durable insect sweep nets easily collected insects
179 from grass, fallow land, brush and the rice crops. Two times 20 complete sweeps were taken
180 to collect insect pests and their NE at maximum tillering stage of the rice crop at each sample
181 plot, because that stage of rice harbors a wider number of arthropods (Bari et al. 2016). The
182 collected insect pests and NE were sorted, identified, counted and written onto a data
183 collection sheet for every sampled plot. Each bund was covered by numerous weed species.
184 The number of weed species grown in the rice bunds was also recorded, but individual weed

185 species were not identified taxonomically, but instead to only estimate the total number of
186 weed diversity, as found on that bund. The width of each rice bund was also recorded using a
187 measuring scale. The plot used to collect arthropods using sweep net was also examined by
188 rice hills (for a total of 100 hills/sampled rice plot), to make additional observations on
189 infestations by insect pests that stay in the lower part of the plant. The sampled insect pests
190 observed in the tested plots in each landscape were negligible and avoid to present in the
191 paper. Therefore only NE populations were described here.

192 Relative abundance of NE populations was calculated using the following equation
193 (Rahman et al. 2014):

$$\text{Relative abundance (\%)} = \frac{\text{Total No. of individuals of each species}}{\text{Total No. of individuals of all species}} \times 100$$

194
195
196 Data were analyzed by means of analyses of variance (one-way ANOVA), using landscape
197 category as the explanatory variable. Means were compared by Tukey's test among the
198 landscapes ($P < 0.05$). The paired t -test was also performed to analyze the effect between rice
199 plot and bund. Data were transferred to logarithm scale in order to homogenize the variance.
200 Pearson's correlation analysis was used to determine the correlation coefficients between NE
201 populations with rice bund's width and number of weed species growing on the bund. All
202 statistical analyses were done using SPSS software, Version 16.0.

203 A LANDSAT 8, true color satellite image in TIFF format, (labelled,
204 LC08_L1TP_137044_20161130_20170317_01_T1) for late November, 2016, was obtained
205 from the United States Geological Survey (USGS) website for variography analyses. It was
206 hypothesized that variograms at the Union area extent of the 6 sampled landscapes and 12
207 more non-sampled, non-classed Unions for comparison, could be useful to aid in landscape
208 classifications of the other Unions of the Barisal and Jhalkerti Districts of the Barisal
209 Division of southern Bangladesh. The software package ERDAS IMAGINE® 2016 and ESRI

210 ARCGIS® 10.2, and SAS® software for PROC VARIOGRAM were utilized to process and
211 analyze the RED spectral pixels contained within the polygon feature layer of the selected
212 Unions (Fig. 1). This date was selected for its cloud free characteristics and because it was
213 not closely associated with the sampling times of the plot survey data, in order to examine
214 how non-seasonal information from satellite platforms may aid the ground survey results and
215 characterization of the sampled landscape classes.

216 **Results**

217 We have assessed four (4) different insect predators under the present study namely, spiders
218 — a general predators group, the green mirid bug (GMB, *Cyrtorhinus lividipennis* Reuter) —
219 an egg predator of planthoppers and leafhoppers of rice, the carabid beetles (CB) — predators
220 of several kinds of planthoppers and leaffolder larvae of rice, and the staphylinid beetles
221 (staphilinids) — a generalist predators of the nymphs of planthoppers. The number of insect
222 pest species found was negligible during this phenological stage while sampling NE, having
223 no effect on rice yields. The sampled plots (and plots) were also tracked up to harvesting
224 stage in order to detect if a significant number of pests occurred after data recording. The
225 investigated plots did not show any visual plot damage due to insect pests. But the number of
226 NE population varied among the landscapes and their population dynamic was described
227 below.

228 **Spider:** Spiders were significantly different among the landscape categories recorded from
229 the landscape I, landscape II and landscape III rice plots. The landscape I showed the highest
230 population numbers than that of the other two categories (Fig. 2). Number were significantly
231 higher in the rice (plot) plots than the rice bund in landscape I and landscape II, but the
232 highest population was found in rice bunds in the landscape III (Fg. 2; $p = 0.001$). The spider
233 populations also depended on the width of rice bund and a trend of increase followed the
234 increase with bund width (Fig. 3). A significant correlation existed between spiders and bund

235 width (Pearson's correlation, $r = 0.576$; $p = 0.050$), but there was no correlation between
236 spider numbers and the number of weed species found in the bunds (Pearson's correlation, r
237 = -0.119 ; $p = 0.80$). The relative abundance of spider population also significantly differed
238 among the different landscapes (Fig. 7).

239 **Green mirid bug (GMB):** Population of GMB was significantly different among the
240 landscapes. Rice plots located in landscape I showed the highest populations compared to that
241 of the other two (Fig. 4, $df = 9$, $F = 167.58$, $p < 0.001$), while the lowest population was
242 observed in the rice plots of landscape III. Similar number of GMB population was found in
243 the rice bund of both landscape I and landscape III and significantly less individuals were
244 observed on the rice bunds of landscape II (Fig. 4, $df = 9$, $p < 0.05$). Populations of GMB
245 differed significantly between rice plots and rice bund among all landscapes ($df = 9$, $p <$
246 0.001). The relative abundance of GMB population also significantly differed among the
247 different landscapes (Fig. 7).

248 **Carabid beetles (CBB):** Populations of CBB significantly varied among all landscapes,
249 where landscape III showed the highest numbers in the rice plots than that of the other two
250 landscapes (Fig. 5, $df = 9$, $F = 167.58$, $p < 0.001$), with the least found in landscape II. In
251 regards to the rice bund, highest number of CBB population was found in landscape III and
252 the lowest population was observed in landscape I. The CBB population significantly varied
253 between rice plots and rice bund at every landscape ($df = 9$, $p < 0.001$). Like the spiders, the
254 CBB abundance also depended on the width of rice bund and increased with increased bund
255 widths. Significant correlation existed between these two variables (Pearson's correlation, $r =$
256 0.423 ; $p = 0.050$). No significant correlation existed between CBB numbers and number of
257 weed species found on the bund (Pearson's correlation, $r = 0.119$; $p = 0.78$). The relative
258 abundance of CBB also significantly differed among the different landscapes (Fig. 7).

259 **Staphylinid beetle:** Landscape III contained significantly more staphylinids than landscapes
260 I and II. (Fig. 6, $df = 9$, $p < 0.05$). The rice bund located in landscape III harbored the highest
261 number of staphylinids where the lowest population numbers occurred in Rotab. The
262 staphylinids varied significantly between rice plots and rice bund in landscape I ($df = 9$, $p <$
263 0.05) but did not show so in the other two landscapes ($p > 0.05$). Staphylinid abundance in
264 the rice bund did not depend on the width of the rice bund. No significant correlation existed
265 between these two variables (Pearson's correlation, $r = 0.322$; $p = 0.481$). Also, no
266 correlation existed between staphylinids and the number of weed species grown in rice bund
267 (Pearson's correlation, $r = 0.243$; $p = 0.599$). The relative abundance of staphylinids also
268 significantly differed among the different landscapes (Fig. 7).

269 **Variography Analyses.** Variography is a typical geostatistical procedure used in analyses of
270 imagery (Stein *et al.*, 2002). For the purposes of this research, however, the variograms
271 derived by each Union, subsetting out only the RED pixel attribute for estimating the
272 empirical variogram, used a RANGE parameter large enough to span the breadth of each
273 Union, regardless of its size and shape. Hence, the variograms plotted show more undulation
274 than most often found by other analysts, whom only seek to estimate a range distance where
275 the SILL parameter reaches its initial plateau (Fig. 8). To understand how variography of a
276 large areal extent can relate to landscape level characterizations, more liberty in the range
277 attribute of each variogram of each of the six Unions was granted. Each union is designed as
278 one landscape based on their surrounding characteristics. Variogram analysis on remote
279 sensing data determine spatial heterogeneity, providing insight into interesting spatial
280 characteristics of the crops and could be extend to multitemporal analyses (Zhaofei *et al.*
281 2013; Duveiller and Defourny 2010; Levin 1992).

282 **Discussion**

283 The presence of NE is a vital component of Integrated Pest Management (IPM) for
284 control of the insect pests of any crop. In this sampling experiment we tried to quantify the

285 natural enemy's abundance prevailing in three distinct rice landscapes. We found that the
286 natural enemy's numbers varied significantly among the sampled landscapes. We took
287 samples (Wang et al. 2015) from the three different locations and showed significant
288 differences in natural enemy's abundance among the landscapes, and also, most times,
289 between those found in the rice plots compared to the rice bund. We hypothesize that the
290 variation in NE abundance among these landscapes arose due to environmental factors (such
291 as, temperature, landscape elevations, historical plant communities, soil characteristics,
292 weather forcings, etc) or anthropogenic factors (that is, cultural or pest management
293 practices). At this stage of investigation, we cannot consider which one of a list of factors
294 most influences the results of this study.

295 High abundance of GMB population was found in rice plots in landscape I located in
296 Protap ($p < 0.01$), but the host insect pest prey for GMBs were not found in those rice plots.
297 The actual mechanism is unknown to explain why GMB population was so high in landscape
298 I. Because preys of the GMB were absent in sampling plots though GMB population depends
299 on prey populations such as BPH, WBPH etc. We also investigated prey individual using
300 both sweeping and individual rice plants examined by naked eye (visual counting) to identify
301 the prey numbers possibly found in the rice plots. Prey insect pests, especially the brown
302 planthopper (BPH), the white-backed planthopper (WBPH), or the green leafhopper (GLH)
303 were not found in any abundance among the landscapes. Generally, GMB eats eggs of BPH,
304 WBPH, GLH laid in rice stems in plot plots. Here, we can assume that prey population could
305 be very small due to huge number GMB population and prey numbers were so low that we
306 could not measure damage to any rice plant at the sample number sizes we used. However
307 variation of predator's populations can be explained by other ways. We hypothesize that
308 landscape structures might provide the suitable hosting for specific predators that could
309 induce higher number of predators in rice plot. Landscape-structures containing perennial
310 habitats can support higher abundance of NE (Andow, 1991; Schmidt and Tscharntke, 2005)
311 or natural enemy's population can be influenced by that vegetational diversity (Werling et al.,
312 2011).

313 In our study landscape I contained abundant perennial habitat surrounded the rice plot (see
314 methodology section). Therefore, landscape I showed higher number of GMB due to its
315 abundant perennial habitat/ higher vegetational diversity (Bianchi et al., 2006; Werling and
316 Gratton, 2008). A landscape may have a large proportion of semi-natural habitat and be
317 otherwise dominated by a single semi-natural land-cover type such as forest. In this
318 experiment, the presence of semi-natural habitat/ abundant perennial habitat was a better

319 predictor of GMB abundance than habitat diversity (Bianchi et al. 2006). Moreover, we
320 proposed that aquatic weeds growing in narrow canal close to the landscape I rice plot
321 harbored the mirid bug population. Pesticide also influences the natural enemies in rice field.
322 However, all tested landscapes plots received similar amount of pesticide during the crop
323 growth stage. Therefore, we can not explain that pesticide influenced the NE at tested rice
324 plot.

325 Similarly, the highest number of spiders was found in Landscape I. Such may be due
326 to the effect of landscape composition/ecosystem, including the surrounding environmental
327 conditions. Higher numbers spiders were found in rice (plot) plots than the rice bund in both
328 landscape I and landscape II, but in contrast lower numbers occurred in rice plots than the
329 rice bund at the landscape III. Since the population numbers depended on rice bund width
330 (Fig. 3), and the rice bunds were smaller in landscape I and landscape II (approx. 25cm) than
331 in landscape III (75-150cm, or 3-6 times wider), bund margins may be particularly
332 important as a source of colonization by ground dispersing predators, such as large
333 *Pardosa pseudoannulata* spiderlings and adults (Sigsgaard 2000). Wider space of rice bund
334 host greater number of weeds species which too can induce higher numbers of spiders in
335 landscape III bunds. But in our study we did not find any significant correlation between NE'
336 populations and number of weed species grown in rice bund ($df = 8, p > 0.05$), so again,
337 understanding of the mechanisms involved is naïve. More investigations are going to be
338 necessary to increase understanding them.

339 The population of carabid beetles and staphylinid was higher at landscape III than
340 Landscape I located at Protap. The variation might occur once again due to the effect of
341 landscape characteristics, including surrounding environmental conditions or man-made
342 traditions. Agricultural landscape producing existing and introducing new crops for
343 mitigating nutritional secure and food security for additional people. These changes influence
344 predators at local scales, as some crops might provide suitable/or unsuitable habitat than
345 previous one (Maredia et al., 1992a; Bommarco, 1999; Landis et al., 2000).

346 We also applied variogram model for describing the spatial relationship between
347 neighboring locations which are the critical elements of any spatial distribution (Hershey
348 2000). The model is applied to match closely the spatial relationship observed in the sample
349 data. In this report, the %population/plot estimates of four predatory insects were further
350 refined. Spider and mirid species, carabid beetle and staphylinid beetle varied by region, and
351 the process of variogram modeling and estimation was repeated for each region. With mirid
352 bug population, the region was applied to target more specifically those areas in which the
353 species was more dominant. More too needs to be developed with respect to applications of
354 variography, because our initial findings (Fig. 8) suggest that sets of graph pattern and shape
355 can be derived. The need is for further collaboration between the expertise of the Bangladesh
356 rice and Mid-South, USA cotton entomologists do further work of analyses of public domain
357 imagery, not only over years, but of months within years, and different areal extents while
358 also obtaining important count data of NE and pest abundances by on-the-ground sampling
359 work. In addition to these kinds of data layers that relate the temporal resolution of the
360 information, care must also be given to spatial and spectral resolutions of the remote sensing
361 layers as available. Our early collaborations show opportunity for learning more about our
362 various agriculturally distinct landscapes (and arthropod fauna of interest) to eventually
363 accomplish the objective of this research.

364 Conclusion

365 In this study, landscape characteristics affected rice pest NE rather than prey population
366 numbers. Surprisingly, we did not find any prey species of a most important predator, the
367 GMB. At a local scale, NE' abundance increased within the managed habitats. Thus, among
368 the different landscapes they were more often differently captured between the rice plots of
369 plots than the plot edge habitat of the rice bund. In contrast, overall NE' levels in equivalent
370 habitats may be related with proportional abundance of semi-natural, dominating, single

371 habitat types found different within each landscape category. Predator individuals in rice
372 plots to landscape-scale semi-natural habitat was identified by the presence of adjacent
373 environmental components, indicating that for these predators, landscape characteristics
374 override the effect of enhanced local resources. This suggests that to manage for increased
375 bio-control services for rice insect pests will require a focus on manipulating overall
376 landscape structure. A greater understanding of these complex relationships will enable
377 growers and researchers to develop more effective management systems suited to specific
378 landscapes, prevailing pests, and their natural enemy communities. Thus, we may anticipate
379 that in the future a combination of local and landscape management practices may be
380 required to maximize overall pest suppression in these larger agro-ecosystems.

381

382 **Conflict of interest**

383 Authors declare that they have no conflict of interest.

384 **Acknowledgement**

385 We greatly appreciate Dr JF Willers for developing variogram model and preparation of
386 figure 8 reported in this paper. We also acknowledge those who (plot staff/assistants) helped
387 to record data and plot experiment and especially we acknowledge farmers for providing land
388 for this experiment. The authors would like to extend their sincere appreciation to the
389 Deanship of Scientific Research at king Saud University for its funding this Research Group
390 NO (RG-1435-014).

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398 **Figure caption**

399 Fig. 1: Map of the Southern part of Bangladesh, for the Barisal Division, where six unions
400 contained the three landscape categories (see text for listing) sampled for insects, while all
401 eighteen unions were employed for variography analyses.

402 Fig. 2: Abundance of spider in three rice landscapes. Means followed by the same letter do
403 not differ significantly at 5% level. The error bar represents the standard error. (***, ** and *
404 indicate significant at 0.1%, 1% and 5% level respectively. Capital and small letters indicate
405 the rice bund and rice plot in different landscapes respectively.)

406 Fig. 3: Effect of rice bund width on the spider individuals. Ten complete sweeps were used to
407 record population from rice bund.

408 Fig. 4: Abundance of green mirid bug (GMB) in three rice landscapes. Means followed by
409 the same letter do not differ significantly at 5% level. The error bar represents the standard
410 error. (*** indicates significant at 0.1% level. Capital and small letters indicate the rice bund
411 and rice plot in different landscapes respectively.)

412 Fig. 5: Abundance of carabid beetles (CDB) in three rice landscapes. Means followed by the
413 same letter do not differ significantly at 5% level. The error bar represents the standard error.
414 (*** indicates significant at 0.1%, level. Capital and small letters indicate the rice bund and
415 rice plot in different landscapes respectively.)

416 Fig. 6: Abundance of staphylinid beetles (STPD) in three rice landscapes. Means followed by
417 the same letter do not differ significantly at 5% level. The error bar represents the standard
418 error. (* indicates significant at 5% level. Capital and small letters indicate the rice bund and
419 rice plot in different landscapes respectively. ns: non-significant at 5% level.)

420 Fig. 7: Relative abundance of four predators in three rice landscapes.

421 Fig. 8: Empirical variograms in one direction for the RED spectral band of an LANDSAT 8
422 image of 30 m ground spatial distance per pixel, of 6 southern Bangladesh Unions, were
423 sampled for NE abundances.

424 **Reference**

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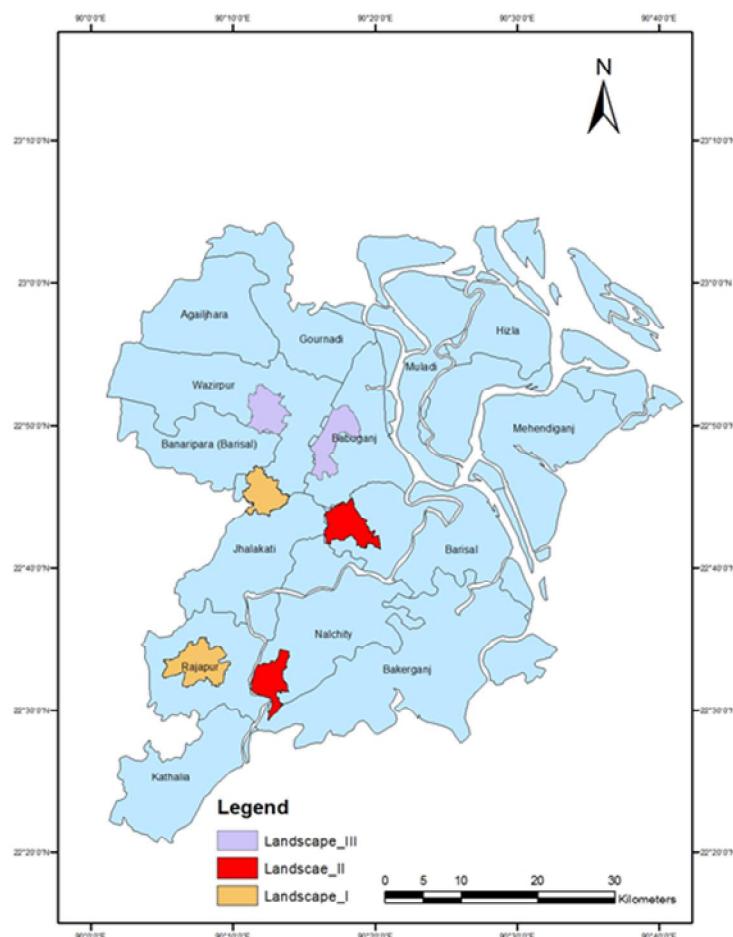
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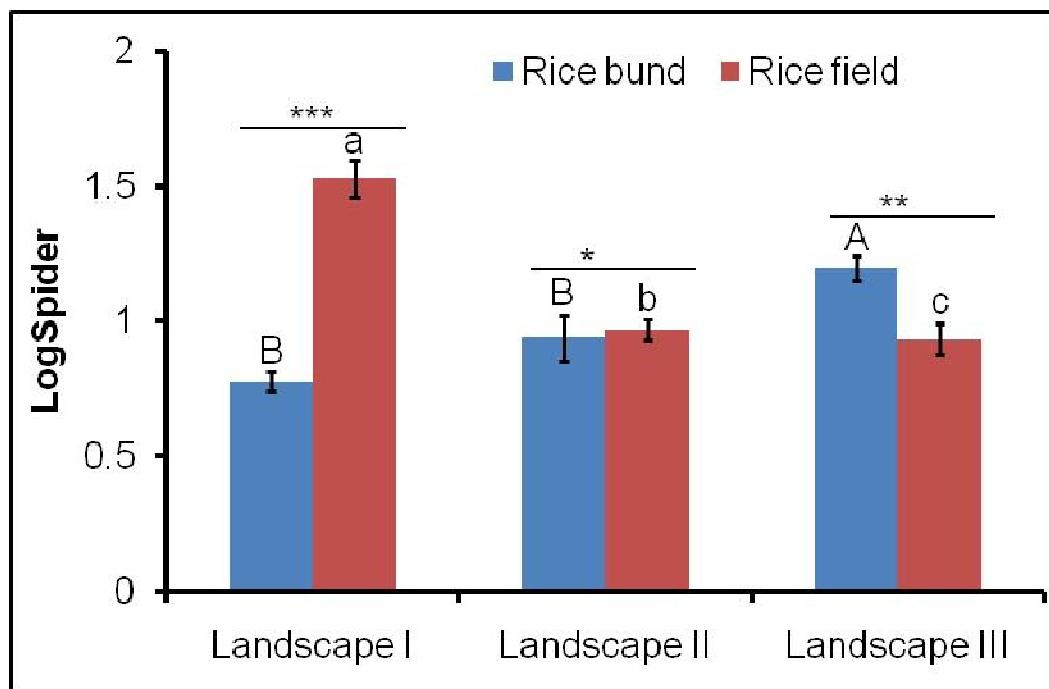
522 Fig. 1: Map of the Southern part of Bangladesh, for the Barisal Division, where six unions
523 contained the three landscape categories (see text for listing) sampled for insects, while all
524 eighteen unions were employed for variography analyses.



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526 Fig. 2: Abundance of spider in three rice landscapes. Means followed by the same letter do
527 not differ significantly at 5% level. The error bar represents the standard error. (***, ** and *
528 indicate significant at 0.1%, 1% and 5% level respectively. Capital and small letters indicate
529 the rice bund and rice plot in different landscapes respectively.)

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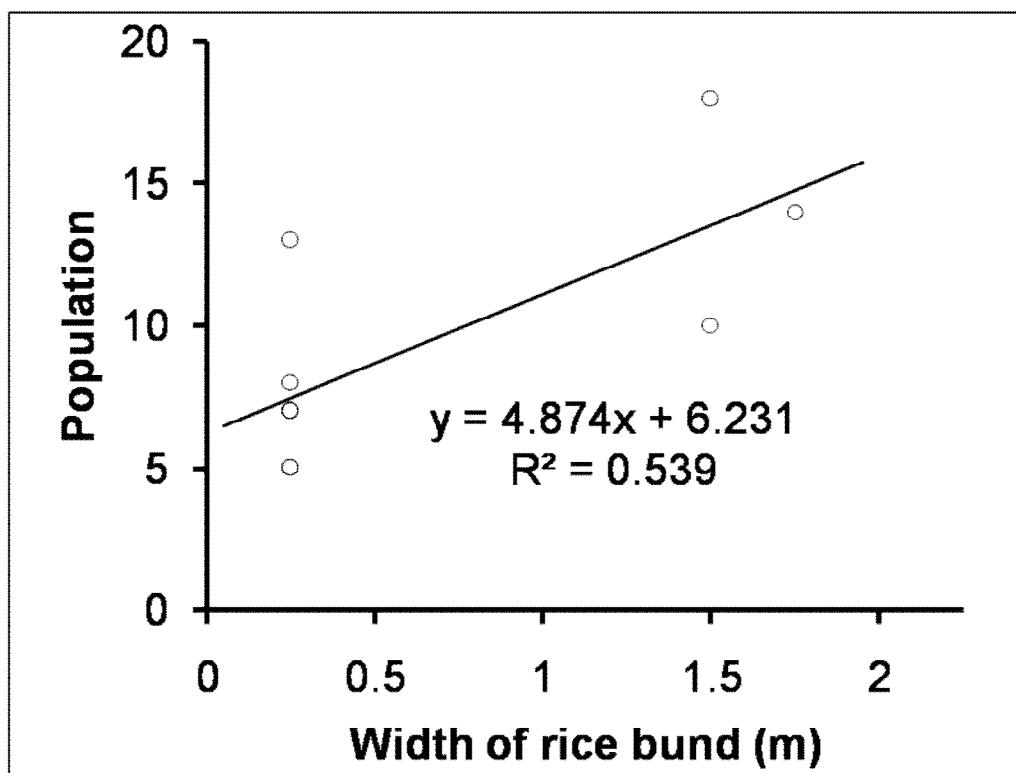
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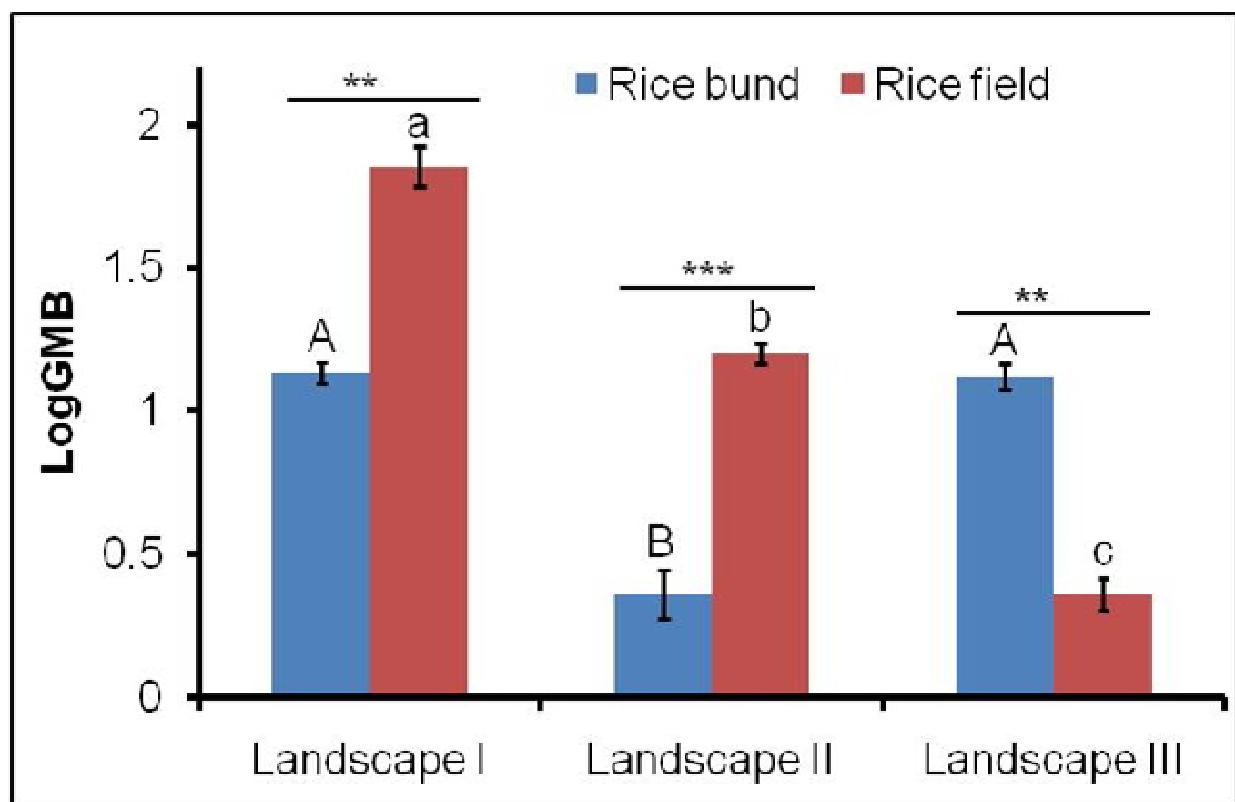
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539 Fig. 3: Effect of rice bund width on the spider individuals. Ten complete sweeps were used to
540 record population from rice bund.



551 Fig. 4: Abundance of green mirid bug (GMB) in three rice landscapes. Means followed by
552 the same letter do not differ significantly at 5% level. The error bar represents the standard
553 error. (*** indicates significant at 0.1% level. Capital and small letters indicate the rice bund
554 and rice plot in different landscapes respectively.)

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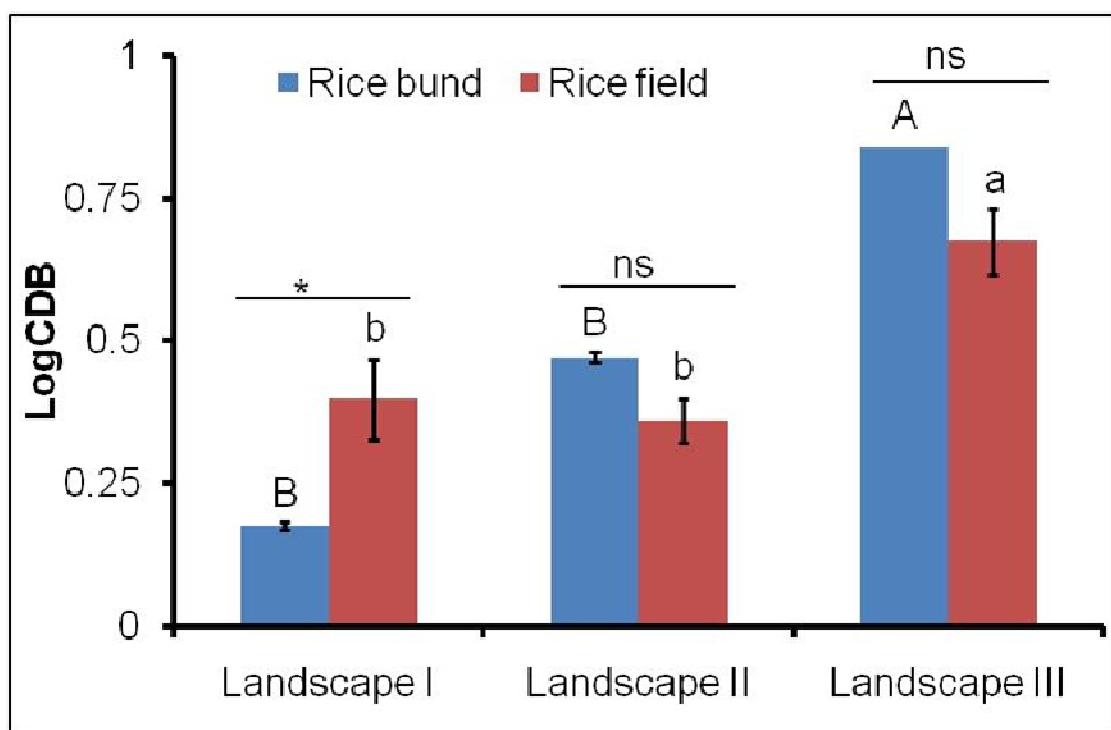
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564 Fig. 5: Abundance of carabid beetles (CDB) in three rice landscapes. Means followed by the
565 same letter do not differ significantly at 5% level. The error bar represents the standard error.
566 (** indicates significant at 0.1% level. Capital and small letters indicate the rice bund and
567 rice plot in different landscapes respectively.)



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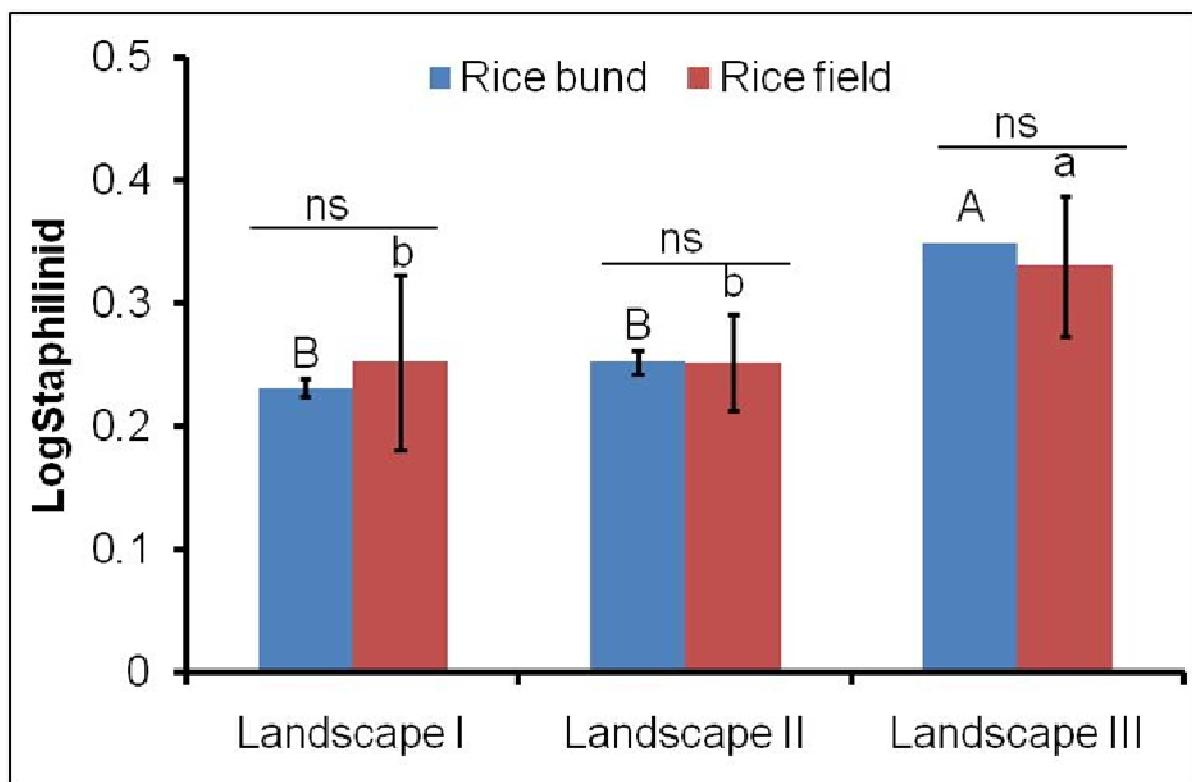
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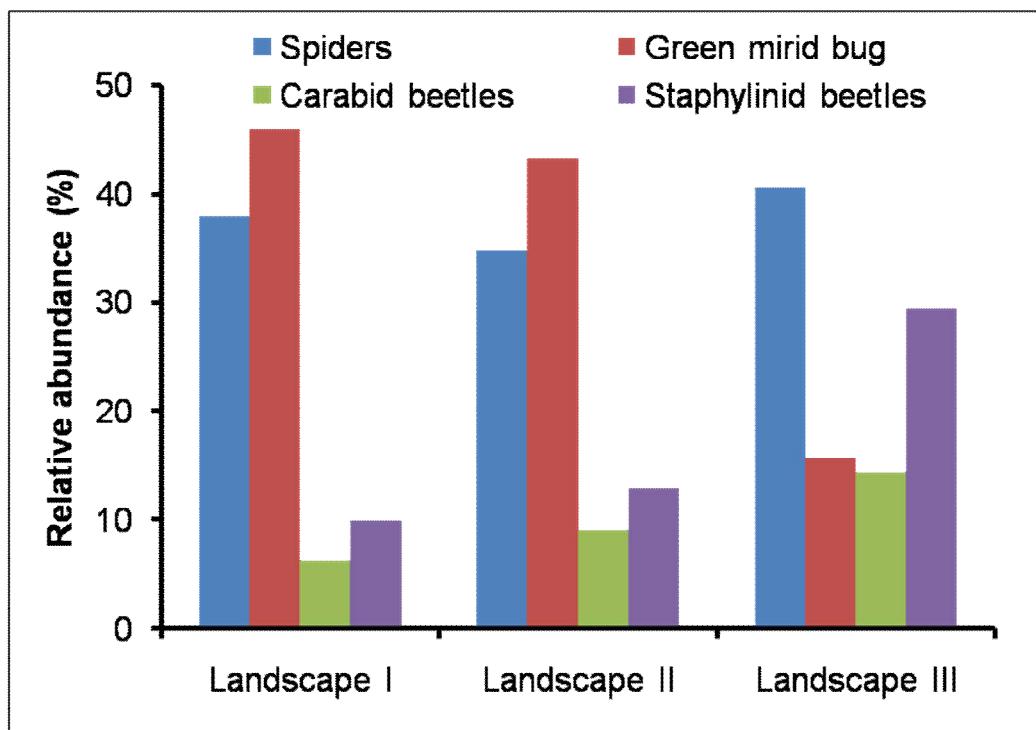
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577 Fig. 6: Abundance of staphylinid beetles (STPD) in three rice landscapes. Means followed by
578 the same letter do not differ significantly at 5% level. The error bar represents the standard
579 error. (* indicates significant at 5% level. Capital and small letters indicate the rice bund and
580 rice plot in different landscapes respectively. ns: non-significant at 5% level.)



590 Fig. 7: Relative abundance of four predators in three rice landscapes.



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601 Fig. 8: Empirical variograms in four directions for the RED spectral band of an LANDSAT 8
602 image of 30 m ground spatial distance per pixel, of 6 southern Bangladesh Unions, which
603 were sampled for NE abundances.

