# Journal of Cereal Research

13(2): 171-179

Homepage: http://epubs.icar.org.in/ejournal/index.php/JWR

# Additive Main effect and Multiplicative Interaction (AMMI) model analysis for yield performance and G×E interaction in a multi-environmental trial of aromatic fine rice in Bangladesh

Shams Shaila Islam<sup>1\*</sup>, Ahmed Khairul Hasan<sup>2</sup>, Abul Bashar Mohammad Khaldun<sup>3</sup> and Novizar Nazir<sup>4</sup>

<sup>1</sup>Department of Agronomy, Faculty of Agriculture, Hajee Mohammad Danesh Science and Technology University, Dinajpur 5200, Bangladesh

<sup>2</sup>Department of Agronomy, Faculty of Agriculture, Bangladesh Agricultural University, Mymensigh-2202, Bangladesh <sup>3</sup>Oil Research Centre, Bangladesh Agricultural Research Institute, Gazipur-1701, Bangladesh <sup>4</sup>Department of Quality control and Agro Inductrial Management, Faculty of Agricultural Technology, Andalas

<sup>4</sup>Department of Quality control and Agro Industrial Management, Faculty of Agricultural Technology, Andalas University-25163, Indonesia

### Article history:

Received: 4 July, 2021 Revised: 8 Aug., 2021 Accepted: 21 Aug., 2021

### Citation:

Islam SS, AK Hasan, ABM Khaldun and N Nazir. 2021. Additive Main effect and Multiplicative Interaction (AMMI) model analysis for yield performance and G×E interaction in a multi-environmental trial of aromatic fine rice in Bangladesh. *Journal* of *Cereal Research* **13(2)**:171-179. <u>http://doi.</u> org/10.25174/2582-2675/2021/114189

\*Corresponding author: E-mail: shaila.hmdstu@gmail.com & shaila@hstu.ac.bd

© Society for Advancement of Wheat and Barley Research

### Abstract

The problem of genotype-environment interaction  $(G \times E)$  in interpreting multilocus trial analyses and predicting genotype performance can be mitigated by applying Additive Main effect and Multiplicative Interaction (AMMI) model analysis. The model AMMI was used in the present study to determine the effect of genotype, environment, and their interaction, to determine the extent of G×E interaction and to identify the factors contributing to G×E interaction for the best yielding aromatic fine rice genotype grown in four different districts in Bangladesh. Analysis of variance showed that the effect of genotypes, environments, G×E interactions were highly significant for plant height, days to maturity, panicles length, grain yield. Result showed grain yield had highly significant differences for environmental traits like (soil properties, phenological, genotypes) with their interactions which indicated that environments were different and changeability with the genotypes. Large variations in total P, Fe and rainfall identified as the main cause of the observed interaction. Here, Genotype BRRI Dhan34 had the highest mean grain yield values over four locations, respectively. In AMMI model, among four locations Dinajpur with Nilphamari were the majority responsive environments and Dinajpur was the most adjacent responsive location. Therefore, location Dinajpur with BRRI Dhan34 genotype could be considered as a better combination for higher grain yield among the ten aromatic fine rice genotypes.

**Key Words:** AMMI model; G × E interaction; grain yield; PCA; aromatic fine rice

# 1. Introduction

Rice (*Oryza sativa* L.) is one of the most important staple foods for more than half of the world's population with a global production of more than 700 million tons per year area of 165 million hectares. (Nayak *et al.*, 2019). Local genotypes, including aromatic fine rice, occupied about 12.16% of the total rice area in Bangladesh (Akter *et al.*,

2020). All rice genotypes have different specialty while some are very prevalent for their aroma and scent. Many countries such as India, Thailand, Vietnam, USA, China, etc., are involved in developing special aromatic rice cultivars (Verma *et al.*, 2018). Although the productivity of aromatic fine rice is comparatively low, its demand for



**Research** Article

internal consumption for export is increasing gradually (Haque *et al.*, 2012).

In agricultural research testing a numerous genotype in several environments is called multi-environment trials (MET). MET is usually conducted to find superior genotypes for better cultivation in the future (Diyah and Hadi, 2016). A variety or genotype is more adaptable if it gives high average yield but has low variation in yielding ability when grown in different environments (Karim *et al.*, 2012). The use of genotype main effect (G) plus genotype-by-environment (GE) interaction (G+GE) AMMI analysis by plant breeders and other agricultural researchers has increased dramatically during the past 5 years for analyzing multi-environment trial (MET) data reported by (Yan *et al.*, 2007).

AMMI model has been expansively applied in the statistical analysis since large portion of the  $G \times E$  interaction sum of squares visibly separates main and interaction effects which support a breeding program over the check locations (Ebdon and Gauch, 2013 and Rodrigues *et al.*, 2014). Therefore, estimation GEI by AMMI model is the best method stated by (Kindeya *et al.*, 2015).

AMMI model combines ANOVA for the G×E effects with the additive parameters of principal component analysis (PCA) reported by (Gauch and Zobel, 2006). Thillainathan and Femandez (2001) designated that the biplot display of PCA scores plotted against each other provides visual inspection and interaction components. Application of the AMMI model has performed normally throughout the previous two aeras namely (Eberhart and Russell, 1966), variance component methods (Shukla 1972; Gauch *et al.*, 2008; Yang *et al.*, 2009; Rodrigues *et al.*, 2014). With this background, the main objective of the current study was to identify the aromatic rice; perform G x E interaction and find out the influence of environmental components related to G x E interaction to better control the yield of aromatic rice.

### 2. Materials and Methods

### 2.1 Experimental plant materials

The experimental plant materials *i.e.*, ten aromatic fine rice genotypes (Table 1) collected from different district of Bangladesh. Four locations differing in latitude, longitude and elevation from the sea level were in Dinajpur (25°37'38" N, 88°38'16" E and 42 m); in Thakurgaon (26°41'83" N, 88°42'16"E and 60 m); Panchagarh (26°20'00" N, 88°33'27" E and 79 m) and Nilphamari (25°48'27" N, 88°41'27" E and 40 m).Top of FormTop of Form The genotypes were evaluated in a RCBD with three replicates in a plot size of 4m×5m with a spacing of 30 cm between rows. Experiments related soil components were described in (Table 2). For the final setting of the experiment, 30-day-old seedlings were used, and one seedling was transplanted per hill. Adequate soil fertility was ensured by applying urea, Triple Super Phosphate (TSP), muriate of potash (MOP), gypsum, ZnSO4 @ 250:130:120:50:10 kg/ha, respectively. Different agronomic actions namely, weed control overall completed by manually, insect and pests by the solicitation of 20 ml per 1 L Cypermethrin 10% w/v EC and 50 ml per 1 L Benfuracarb 20% w/v EC. At 30 days after planting urea fertilizer (46-0-0) was applied.

Table 1	1:1	Ten 1	popular	aromatic	fine	rice	genotypes	with	place of	collection	in	Bangladesh
---------	-----	-------	---------	----------	------	------	-----------	------	----------	------------	----	------------

Sl No	Genotypes	Place of collection	Kernel size and shape	Yield (T/ha)
G1	Kataribhog	Dinajpur	Short, medium Scented	2.00
G2	Kalijira (medium grain)	BRRI	Short, medium Scented	1.96
G3	Kalijira (long grain)	Khulna	Short, medium Scented	2.11
G4	BRRI dhan34	BRRI	Short, medium Scented	2.66
G5	BRRI dhan37	BRRI	Short, medium Scented	1.92
G6	Chinigura	Sherpur	Medium, slender Scented	1.21
G7	Basmati	Barguna	Short, bold Scented	2.43
G8	Tulsimala	Sherpur	Short, bold Lightly scented	1.95
G9	Badshabhog	Dhaka	Short, bold Scented	1.35
G10	Gobindhabhog	Jessore	Short, medium Lightly scented	2.38

### Journal of Cereal Research 13(2): 171-179

### 2.2 Data collection

Yield contributing characters like plant height, days to maturity, panicles length, grain yield data was recorded using ten randomly selected plants in each replication and yield data were finally converted in to (t/ha).

## 2.3 Statistical analysis

Grain yield which was collected at 12% moisture level. Observations were recorded and the data were statistically analyzed. Here, the contribution of each genotype and environment to GEI is assessed by using the biplot plot in which mean yield values are plotted against scores of the first principal component interaction (PCA1). Correlation coefficient, analysis of variance (ANOVA) with Post Hoc also PCA (Principal Component Analysis) are using SPSS (ver 20) and XLSTAT (ver16).

# 2.4 Additive main effect and multiplicative interaction (AMMI) method for RCBD analysis

The AMMI method was applied with additive effects to 10 genotypes in three environments, and multiplicative was used for  $G \times E$  interaction. It affords a symbolic view of the transformed  $G \times E$  interaction for any interpretation (Kempton, 1984) based on the following AMMI equation:

Where, = Yield for genotype g, environment e and replication r;  $\mu$  = Grand mean value for trait; = Mean deviations for genotype (genotype means minus grand mean); = Mean deviation for environment; n = PCA axis number reserved in the model; = Singular value for PCA axis n; = Genotype eigenvector values for PCA axis n; = Eigenvector for environment; = Residuals and = Error is used

Table 2:	Ph	ysico-	chemical	charac	teristics	of	initial	soils	in	the	different	ex	perime	ental	fiel	ds
----------	----	--------	----------	--------	-----------	----	---------	-------	----	-----	-----------	----	--------	-------	------	----

Soil characteristics	Locations							
	Dinajpur	Thakurgaon	Panchagarh	Nilphamari				
Soil texture	Sandy clay loam	Loomy	Sandy clay loam	Sandy clay loam				
pH	5.25	5.1	5.7	5.21				
Organic carbon (%)	0.81	0.55	0.79	0.79				
Total N (%)	0.08	0.07	0.08	0.08				
Available P (mg/kg soil)	120.34	98.99	121.44	120.00				
Exchangeable K (cmol/kg)	0.151	0.055	0.141	0.171				
Available S (mg/kg soil)	14	24	14	13				

# 3. Results and discussions

# 3.1 Analysis of variance result including the partitioning of the $G \times E$ interaction of aromatic fine rice

The analysis of result showed significant differences for plant height, days to maturity, panicle length and grain yield for genotype, environment, G x E interaction. The highly significant effect on environment indicates high differentiation of genotypic responses in different environments and existence of wide range of diversity among genotypes (Kulsum *et al.*, 2012). Analysis of variance based on the AMMI model for grain yield is shown in Table 3 indicating that genotype performance was more influenced than environmental factors. The genotype × environment effect interaction could be divided into two components, namely IPCA1 and IPCA2. All significant differences were found for grain yield, indicating that the components of G×E interaction affected the yield of genotypes in different environments and the environments were different. Kumar *et al* (2012) reported the result for hybrid rice at different locations in Bangladesh.



		Mean sum of squares							
Source of variation	df	Plant height	Days to maturity	Panicle length	Yield (t/ha)				
Genotype (G)	9	45.22**	69.88**	1.756**	2.22**				
Environment (E)	3	263.33**	533.77*	11.26**	1.80**				
Replication (R)	2	22.01**	4.82**	1.00*	0.065*				
Interaction (G×E) GEI	36	5.033*	2.11**	0.56*	0.142**				
AMMI component 1	12	11.65**	3.44**	0.89*	0.32**				
AMMI component 2	11	5.67**	2.45**	0.76*	0.14**				
Error	18	3.66	0.567	0.578	0.89				

**Table 3:** Analysis of variance including the partitioning of the G×E interaction of aromatic fine rice

Here \* p< 0.05,\*\*p<0.01

3.2 Analysis of variance result for grain yield with soil, climatic and phenological properties

Based on ANOVA result, the average highest grain yield (2.66 t/ha) was found from G4 (BRRI dhan 34) genotype and comparatively low (1.95 t/ha) for G8 (Tulsimala) genotype. High variations occurring in this result were caused by several factors such as soil properties like Fe, total phosphorus, Ca as well as rainfall (Eberhart and Russell, 1966). Changeable environmental features such

as rainfall through a single situation can underscore dissimilarity of genotypes in relation to environment across locations. For the different location trials, the location in which the field trials were undertaken showed geographical and environmental dissimilarities (Islam *et al.*, 2020) .The soil properties showed that the highest variation occurred phenological traits like plant height, days to maturity, panicles length showed highest variation. All the findings shown in (Table 4).

**Table 4:** Summary statistics of grain yield, soil, climatic and phenological properties of ten aromatic fine rice genotypes in 4 locations of Bangladesh

Variables	Sum	Average	Variance	Variables	Sum	Average	Variance
G1	8.012	0.200	0.530	Fe	1068.77	26.71	7138.41
G2	3.800	0.096	0.115	CEC	16.05	0.40	1.521
G3	8.420	0.211	0.578	$_{\rm pH}$	19.95	0.49	2.297
G4	10.63	0.266	0.972	EC	93.76	2.34	52.197
G5	7.68	0.192	0.504	T_max	142	3.55	116.36
G6	4.84	0.121	0.187	T_min	99	2.48	56.61
G7	9.70	0.243	0.681	Rainfall	465.8	11.65	1254.83
G8	3.81	0.095	0.121	Humidity	339	8.48	663.54
G9	5.41	0.135	0.211	PH	523.2	13.08	1622.24
G10	9.51	0.238	0.703	DM	2743.2	68.58	5300.17
Ν	3.78	0.332	0.675	PL	103.6	2.59	63.15
OC	4.44	0.111	0.115	-	-	-	-
Total_P	726.84	18.17	3226.57				
Available_P	29.36	0.73	5.892	-	-	-	-
K	125.02	3.12	91.79	-	-	-	-
Ca	335.15	8.37	664.58	-	-	-	-

Here, N=Total Nitrogen, OM=Organic Matter, OC=Organic Carbon, Total P=Total phosphorus, FG=Filled grains (no), SW=Seed weight (gm), Available P=Available Phosphorus, K=Potassium, Ca=Calcium, Fe=Iron, CEC=Cation Exchange Capacity, pH=pH level, EC=Exchangeable Cation, Tmax=Temperature Maximum, Tmin=Temperature Minimum, PH=Plant Height (cm), PL=Panicles Length (cm) bold letters indicate correlation is significant at the 0.01% level level level level is a significant at the 0.01% level lev



### 3.3 Principal Component Analysis (PCA)

The AMMI biplot provides a visual expression of the relationships between the IPCA1 and IPCA2 with the mean of the genotypes and environments. Principal Component Analysis is a multivariate technique that detects figure arrangements with correspondences and differences between variables set up and arranged in a systematic multivariate system (Islam *et al.*, 2020). Tables 5 and 6 showed that the output of PCA analysis exposed the relation between retained factors and the variables before and after rotation. Figure 1 is the map

titled correlation circle (below on axes PCA1 as well as PCA2) showing a projection of the primary variables into the factors planetary. If two variables were away from the center (Xlstat, 2017) but variables were close to each other then they were significantly positively correlated (r near 1). Besides, remained orthogonal, then they do not exist correlated (r near 0); if they were on the opposite side of the center, then they remained significantly negatively correlated (r near -1). From the plot of the component loadings gave a visual representation schemed in planetary that showed exactly how closely related the items to each other as well as with the components.

Variables	PCA1	PCA2	Variables	PCA1	PCA2
G1	0.725	-0.340	Fe	-0.637	0.753
G2	0.957	-0.073	CEC	0.024	0.977
G3	0.839	0.211	PH	0.598	0.795
G4	0.885	0.299	EC	0.802	0.582
G5	0.876	0.247	T_max	0.652	0.748
G6	0.932	-0.361	T_min	-0.267	-0.944
G7	-0.906	0.421	Rainfall	0.077	0.084
G8	0.903	-0.119	Humidity	-0.174	-0.437
G9	0.958	-0.147	PH	0.977	-0.020
G10	0.974	-0.016	PL	0.971	0.137
Total_N	-0.475	0.853	MD	0.343	0.905
ОМ	-0.449	0.828	-	-	-
OC	-0.281	0.949	-	-	-
Total_P	0.127	-0.108	-	-	-
Available_P	0.779	-0.401	-	-	-
К	0.378	-0.063	-	-	-

-0.232

Table 5: Outputs of PCA analysis between variables and factors before Varimax re
--

In the current study Fig 1(a) the first principal component axis (PCA1) illustrated 55.20% of entire variation while PCA2 explain 26.63%. Therefore, the two axes together explained 81.83% of the G×E interaction for grain yield with other traits and Fig 1(b) also support the same G×E interaction. Figure 1(a) showed that all the variables have strong relationships with some of the environmental parameters before rotation. Therefore, genotypes were closed with their related environmental traits. Table 4 indicated that most of the genotypes with very strong PCA1 values G1 (0.725), G2 (0.957), G3 (0.839), G4

0.841

(0.885), G5 (0.876), G6 (0.932), G8 (0.903), G9 (0.958)and G10 (0.974) appeared with Available\_P (0.779), Ca (0.841), EC (0.802), PH (0.977), PL (0.971) in PCA1. On the other hand, in PCA2 only OM (0.828) and OC (0.949)closed to each other. This finding very much similar with (Poramate and Anchalee, 2015).

Although some of the traits shown strong relationship values but they were not closed to each other. Table 6 showed that after rotation the PCA values between factors and the variables changed to some extent. Figure 1(b) also supported the relationships. The correlation round



Ca

also useful in understanding the significance of the axes. Following this research, the parallel axis link by means of Total\_P, Available\_P, K, Ca, EC, PH, PL, FT and the vertical axis with Total\_N, OM, OC, Fe, CEC, PH, Temp\_max, Temp\_min, Rainfall, Humidity and LI. These trends revealed that a variable is well linked with an axis.



Fig 1: AMMI model based on environments focused scaling for comparison the genotypes with the ideal genotype on grain yield. a) before rotation b after rotation

Variables	PCA1	PCA2	Variables	PCA1	PCA2
G1	0.779	-0.185	Fe	-0.777	0.607
G2	0.652	-0.124	CEC	-0.176	0.961
G3	0.800	0.398	PH	0.424	0.900
G4	0.806	0.473	EC	0.666	0.733
G5	0.807	0.424	T_max	0.486	0.865
G6	0.986	-0.164	T_min	-0.069	-0.979
G7	-0.973	0.228	Rainfall	0.058	0.098
G8	0.909	0.068	Humidity	-0.081	-0.464
G9	0.968	0.051	PH	0.961	0.179
G10	0.751	-0.214	PL	0.922	0.332
Total_N	-0.639	0.738	MD	0.151	0.956
OC	-0.469	0.872	GY	0.949	0.309
Available_P	0.845	-0.234	-	-	-
К	0.383	0.015	-	-	-
Ca	0.871	-0.056	-	-	-

Table 6: Outputs of PCA analysis between variables after Varimax rotation



# 3.4 Analysis of variance result for grain yield in four locations

There had a significant effect of environmental parameters for grain yield with four locations. As shown (Table 7), the Tukey's HSD (Honestly Significantly Different) test smeared to wholly pairwise variances among mean values. The risk of 5% chosen values used to define the critical value F, which compared to the standardized difference between the means. Only three pairs appeared significantly different (E1, E3), (E4, E3) and (E2, E3). The means and the groups then categorized founded on this analysis. In conclusion, four location's different environmental parameters showed significantly effects on the yield of genotypes. Based on the result of Table 7, Fig 2 showed the eventual objective of the Principal Component Analysis. Where E1 and E4 environment nearer to the central point was nearer with similar yield and E1 devoured higher yield than E4. After that E3 and E4 took very low yield comparing to other environments. In this case the best environment was E1 (Dinajpur). It was enabled the observations on a two-dimensional map and to identify links that exists grain yield Dinajpur, Thakurgaon, Panchagarh and Nilphamari locations were unique. It showed that every geographical location had its own environmental characteristics and those characteristics had different impact upon the genotypes.

 Table 7: Environment /Tukey (HSD)/ ANOVA Analysis of the differences between grain yield with four locations in Bangladesh

Contrast	Difference	Standardized difference	Critical value	Pr > Diff	Significant
E1 vs E3	2.061	4.335	2.693	0.001	Yes
E1 vs E2	0.670	1.409	2.693	0.502	No
E1 vs E4	0.403	0.848	2.693	0.831	No
E4 vs E3	1.658	3.487	2.693	0.007	Yes
E4 vs E2	0.267	0.562	2.693	0.943	No
E2 vs E3	1.391	2.925	2.693	0.029	Yes
	Tukey's d cri	tical value:	2.868		



Fig 2: AMMI observation study for comparison of the locations a) before rotation b) after rotation

#### AMMI and $G \times E$ interaction in aromatic fine rice

### 4. Conclusion

In summary, the statistical model AMMI was used to determine the G×E interaction pattern of grain yield of ten promising aromatic fine rice cultivars. ANOVA showed that significant differences among genotypes, soil properties and phenological traits indicated the presence of large variability among genotypes and locations for yield, with G4 (BRRI Dhan34) and G7 (Tulshimala) being the first and second high yielding genotypes corresponding to environments E1, E2, E3 and E4. The maximum yielding genotypes G4, G7, G10, G3, G1, G5 and G6, G2, G8 were the low yielding genotypes with wide adaptation. The analysis of the four environments showed that there were significant differences between E1 vs. E3, E3 vs. E4, and E2 vs. E3. PCA analysis also showed that location E1 (Dinajpur) was found to be optimum selection site for identification of broad and adaptive genotype of aromatic rice and for other improvement work on aromatic fine rice.

### Acknowledgments

This research work was sponsored by IRT (Institute of Research and Training) Center, project code (5921) of Hajee Mohammad Daesh Science and Technology University Dinajpur, Bangladesh for funding the research and providing all necessary support. The authors would like to thank Agronomy Department, Agriculture Faculty for kindly providing the aromatic fine rice seeds

## Compliance with ethical standards

NA

# **Conflict of Interest**

Authors declare that they have no conflict of interest

### Author contributions

Conceptualization of research and designing of experiments (SSI, ), Conduction of experiment (SSI, AKH, ABMK), Preparation of manuscript (SSI, AKH, ABMK,NN).

### 5. References

- Akter S, FA Bonni, ME Haq, N Shith, N Sultana, MJ Runia, A Siddika and MB Nahar. 2020. Growth and Yield of Traditional Aromatic Rice Cultivars in Boro Season. *Asian Journal of Research in Botany* 3(3): 18-27.
- 2. Diyah HS and AF Hadi. 2016. AMMI Model for Yield Estimation in Multi-Environment Trials: A Comparison to BLUP. Agriculture and Agricultural

Science Procedia. 9 (2016): 163–169. doi: 10.1016/j. aaspro.2016.02.113.

- Ebdon, JS and HG Gauch. 2013. Additive main effect and multiplicative interaction analysis of national turf grass performance trials. II. Cultivar recommendations. *Crop Science*. 42(2013): 497-506.
- Eberhart SA and WA Russell. 1966. Stability parameters for comparing varieties. *Crop Science* 6(2): 36 40.
- Gauch, HG and RW Zobel. 2006. AMMI analyses of yield trials.p.85-122. In Kang MS, and H.G. Gauch (eds). Genotype by environment interaction. CRC, Boca Raton, Florida, USA.
- Gauch HG, H Piepho and P Annicchiarico. 2008. Statistical analysis of yield trials by AMMI and GGE: further considerations. *Crop Science*. 48(3): 866-889.
- Kang MS. 1993. Simultaneous selection for yield and stability in crop performance trials: Consequences for growers. *Agronomy Journal* 85: 754-757.
- Haque MA, MAH Khan, ME Haque, MS Islam and MF Islam. 2012. Effect of nitrogen application on morphological characters and yield attributes of fine aman rice cv. Kalizira. *Journal of Agroforestry and Environment.* 6(1): 67-70.
- Islam, SS, J Anothai, C Nualsri and W Soonsuwon. 2020. Analysis of genotype-environment interaction and yield stability of Thai upland rice (Oryza sativa L.) genotypes using AMMI model. *Australian Journal of Crop Science*. 14(02):362-370. doi: 10.21475/ ajcs.20.14.02. p 1847.
- Kindeya, YB, F Mekbib and EA Abraha. 2020. AMMI and GGE bi-plot analysis for seed yield and oil content of sesame (*Sesamum indicum* L.) genotypes in Tigray, Northern. *Ethiopia Journal of Agricultural Crops.* 6(64):58–67. DOI: https://doi.org/10.32861/ jac.64.58.67
- Kulsum, MU, U Sarker, MA Karim and MAK Mian. 2012. Additive Main Effects and Multiplicative Interaction (AMMI) Analysis for Yield of Hybrid Rice in Bangladesh. *Tropical Agriculture and Development*. 56(2): 53-61.
- Kumar, LB, M Nagaraju and ON Singh. 2012. Genotype × Environment interaction and stability

#### Journal of Cereal Research 13(2): 171-179

analysis of lowland rice genotype. *Journal of Agricultural Science*. **57**(1): 1-8.

- Leard Statistics. 2018. principal-components-analysis-PCA-using-SPSS-statistics. Retrieved from https:// statistics.laerd.com/spss.
- 14. Nayak AK, M Shahid, AD Nayak, B Dhal, KC Moharana, B Mondal, R Tripathi, SD Mohapatra, P Bhattacharyya, N Jambhulkar, AK Shukla, N Fitton, P Smith and H Pathak. 2019. Assessment of ecosystem services of rice farms in eastern India. Ecological Processes (2019):1-16. https://doi.org/10.1186/s13717-019-0189-1.
- Poramate B and J Anchalee. 2015. Evaluation of black glutinous rice genotypes for stability of Gamma oryzanol and yield in tropical environments. *Turkish Journal of Field Crops.* 20(2): 142-149.
- Rodrigues PC, M Malosetti, HG Gauch and FA Van Eeuwijk. 2014. A weighted AMMI logarithm to study genotype-by-environment interaction and QTL-by-environment interaction. *Journal of Crop Science.* 54(2014): 1-16.
- Sukla GK. 1972. Some statistical aspects of partitioning genotype-environmental components of variability. *Journal of Heredity.* 29(5): 237-245.

- Thillainathan M and GCJ Fernandez. 2001. SAS applications for Tai's stability analysis and AMMI model in genotype × environment interaction (GEI) effects. *Journal of Heredity*. 92(4): 367-371.
- 19. Verma DK, PP Srivastav and A Nasaf. 2018. Aromatic rice from different countries: an overview. Page:1-55.
- 20. Xlstat. 2017. One-way ANOVA and multiple comparisons in Excel tutorial. Retrieved from https:// help.xlstat.com/.../2062232- One- way- ANOVAand- multiple- comparisons- in- Excel-tut.
- Yan W, SK Manjit, BM Kang, W Sheila and LC Paul. 2007. GGE Biplot vs. AMMI Analysis of Genotype-by-Environment Data. *Crop Science*. 47 (2007): 641-653.
- Yang, RC, J Crossa, PL Cornelius and J Burgueno.
  2009. Biplot analysis of genotype × environment interaction: Proceed with caution. *Crop Science*. 50(4): 1564-1576