

# PRINCIPAL COMPONENTS ANALYSIS IN RICE GENOTYPES GROWN OVER SEASONS

Ponsiva S.T.<sup>1</sup>, S. Sabarinathan<sup>2</sup>, N. Senthilkumar<sup>1</sup>, P. Thangavel<sup>1</sup>, K. Manivannan<sup>1</sup> and S. Thirugnanakumar<sup>1\*</sup>

<sup>1\*</sup>Department of Genetics and Plant Breeding, Faculty of Agriculture, Annamalai University, Annamalainagar- 608002, Tamil Nadu, India <sup>2</sup>NRRI, Cuttack, Odissa, India

### Abstract

Data were collected on 50 genotypes of rice grown over two seasons for eleven agronomical traits. Both individual and pooled analyses were performed. Among the traits of interest, high coefficients of variation were observed for total number of tillers per plant, number of productive tillers per plant and total dry matter production per plant in season 1, season 2 as well as in pooled analysis. All the eleven traits were statistically significant when subjected to univariate statistics. Principal component analysis resulted in the first three components account for more than 89 per cent of total variation in season 1; 83 per cent of total variation in season 2 and 86 per cent of total variation in pooled analysis. These results could well be used in by rice breeders to develop high yielding lines and new breeding protocols for rice improvement.

Key words : Rice; coefficient of variation; Principal components analysis.

## Introduction

Rice is considered as an important food crop in India and millions of people depend on it for their livelihood. The development of high yielding lines is a continuous process and its success rests on the proper selection of suitable plants to be utilized in cross breeding programme. The effectiveness of selection depends primarily upon the quantum of genetic variability available in the germplasm. A large number of traits are often observed by rice breeders. Some of which may not be use in the discrimination of the germplasm. In such cases principal component analysis (PCA) may be used to reveal patterns and eliminates redundancy in data sets (Adams, 1995; Amy and Prittes, 1991; Majiand Shaibu, 2012). Principal component analysis was designed by Karl (1901) and Hotelling (1933) indicated that PCA is an exploratory tool, to identify unknown trends in a multi-dimensional data set. PLA is one among many techniques that reduced the data into two dimensions (Smith, 2002; Rao, 1964; Ray Chaudhuri et al., 2000). The present study was undertaken to make a classificatory analysis on the rice genotypes by means of PCA.

#### Materials and Methods

Fifty rice genotypes of different eco-geographic were collected from the Department of Plant Genetic Resources, Centre for Plant Breeding and Genetics, Tamil Nadu Agricultural University, Coimbatore, India. The crop was raised in randomized block design, replicated thrice. Each genotype was grown in two rows plot of size 3 m length. A spacing of 20×15 cm was adopted. The crop was raised during Kuruvai (2018) and Samba (2018). Eleven different agronomic traits were observed. Analysis of variance using descriptive statistics such as mean, standard deviation and coefficient of variation for each of the eleven traits were calculated. In order to identify the patterns of variation in agronomic traits, PCA was conducted, the PCA was computed using the following equation:

```
PCA
```

PC 1  $\underline{\underline{P}}$  S<sub>1</sub> a j ' j Where, PC = Principal component,

alj = Linear coefficient – Eigen vectors,

Statistical analyses were performed with Indo-state,

<sup>\*</sup>Author for correspondence : E-mail: thirugnanakumar1962@gmail.com

licensed at NRRI, Cuttack, Odisa.

## **Results and Discussion**

The analysis of variance for means indicated that the differences among the genotypes were significant for all the 13 traits of interest (Table 1). The result is in corroboration with the findings of Bharadwaj *et al.*, (2001); Kutaiah *et al.*, (1986) and Maji and Shaibu (2012). It indicated the necessity to group the rice genotypes in to divergent clusters. In rice breeding, the earliness to

 Table 1: Analysis of variance.

attain flowering cycle from 18.33 days after transplanting is a good trait that could be utilized for screening drought tolerance genotype.

The PCA was used to eliminate redundancy in the data set. Three principal components accounted for must of the variability observed among the rice germplasm collections from different geographical locations in S1 (Table 2). Principal component 1 accounted for 79 per cent of the morphological variation in the rice germplasm collection and was loaded on days to 50 per cent flowering

	Mean squares								
Traits	Replication df=2		Genotypes df=49			Error df=98			
	<b>S1</b>	S2	Pooled	<b>S1</b>	S2	Pooled	<b>S1</b>	S2	Pooled
			analysis			analysis			analysis
Days to 50% flowering	01.36	2.5466	1.02	1422.49**	1581.01**	1720.05**	0.78	1.1779	0.8721
Plant height (cm)	34.99	57.41*	41.60	1035.72**	957.78**	990.81*	19.55	16.08	14.08
Total number of tillers per plant	34.07	17.76	24.66	168.92**	129.36**	142.71**	16.45	19.96	17.19
Number of productive tillers per plant	06.13	01.67	02.86	88.04**	65.84**	70.34**	04.89	3.24	3.13
Panicle length (cm)	07.06	01.86	0.44	34.50**	27.83**	29.10**	04.08	3.54	2.82
Number of filled seeds/panicle	69.85	06.66	11.11	6695.24**	5974.82**	6267.47**	92.56	244.42	125.75
Length of seeds (mm)	00.03	00.02	0.02	1.90**	2.0112**	1.68**	00.01	0.01	0.01
Breadth of seeds (mm)	0.00009	00.004	0.004	0.2553**	0.4411**	0.33**	0.0059	0.003	0.002
Length/breadth ratio of seeds	0.0096	00.02	0.002	1.1649**	1.2266**	1.19**	0.0146	0.01	0.01
100 Seed weight (g)	0.0552	0.01	0.09**	0.9458**	0.9772**	0.91**	0.0229	0.01	0.009
Total dry matter production/plant (g)	3.5698	13.77	1.52	1713.30**	1041.11**	1312.13**	37.30	14.48	18.77
Seed yield/plant (g)	2.0934	03.43	0.20	725.86	1028.17**	704.56**	8.2583	4.87	4.25
Harvest index (%)	4.0667	16.32	1.03	629.58	522.90**	532.79**	13.2426	10.05	7.59

\*Significant at P=5% ; \*\*Significant at P=1%

 

 Table 2: Eigen value and per cent of total variation and component matrix for the principal component axes in Season 1.

Principal components	1	2	3		
Eigen value	11375.30	844.79	696.63		
Percentage of variance	79.26	5.89	4.85		
Cumulative (%)	79.26	85.15	90.00		
Component matrix					
Days to 50% flowering	0.95103	0.05030	0.04601		
Plant height at maturity (cm)	-0.08376	0.15474	0.41421		
Total number of tillers per plant	-0.09106	0.02970	0.11737		
Number of productive tillers per plant	0.07724	0.01151	0.11987		
Length of primary panicle length (cm)	-0.01535	0.11402	0.10139		
Number of grains per panicle	-0.09321	0.31081	0.68646		
Length of seeds (mm)	0.04531	0.74487	-0.38235		
Breadth of seeds (mm)	-0.00475	-0.32762	-0.06712		
L/B ratio of seeds	0.00506	0.41043	-0.00450		
Weight of hundred seeds (g)	0.05096	-0.04568	-0.14654		
Total dry matter production per plant (g)	0.14562	-0.02672	0.33976		
Seed yield per plant (g)	0.00807	-0.03797	0.04626		
Harvest index (%)	0.19841	-0.16744	0.16401		

and harvest index (%). Principal component 2 accounted for about six per cent of the total morphological variation and was loaded on number of filled grains per panicle, length of seeds and length; breadth ratio of seeds. The remaining variables had weak or no discriminatory power. Thus the most important descriptors were those associated with principal component 1 and 2.

In S2, principal component 1 accounted for about 69 per cent of the morphological variation in the rice germplasm collection and was loaded on days to 50 per cent flowering, length of seeds and harvest index. Principal component 2 account for about nine per cent of the total morphological variation and was loaded on length of seeds and seed yield per plant. The remaining variables had weak or no discriminatory power. Thus, the most descriptors were those associated with principal component 1 and 2 (Table 3).

Table 3: Eigen value	and per cent of total	variation and	component	matrix for
the principal	l component axes ir	Season 2.		

Principal components	1	2	3		
Eigen value	11041.17	1421.89	1163.29		
Percentage of variance	68.53	8.83	7.22		
Cumulative (%)	68.53	77.36	84.58		
Component matrix					
Days to 50% flowering	0.81560	0.08282	0.34377		
Plant height at maturity (cm)	-0.06020	-0.03314	0.27168		
Total number of tillers per plant	-0.10539	-0.03262	0.02208		
Number of productive tillers per plant	-0.05940	-0.04956	0.02535		
Length of primary panicle length (cm)	-0.05761	0.04063	0.04819		
Number of grains per panicle	-0.08784	0.00113	0.23077		
Length of seeds (mm)	0.24153	0.73821	-0.20422		
Breadth of seeds (mm)	0.12894	-0.39333	-0.36101		
L/B ratio of seeds	-0.34143	0.13226	0.22145		
Weight of hundred seeds (g)	0.13451	-0.25562	0.03459		
Total dry matter production per plant (g)	-0.03725	-0.04339	0.63834		
Seed yield per plant (g)	0.05796	0.27878	-0.27870		
Harvest index (%)	0.30155	-0.35322	-0.20979		

**Table 4:** Eigen value and per cent of total variation and component matrix for the principal component axes in pooled analysis.

Principal components	1	2	3		
Eigen value	1178.43	1934.86	1000.25		
Percentage of variance	69.12	11.35	5.87		
Cumulative (%)	69.12	80.47	86.34		
Component matrix					
Days to 50% flowering	0.90457	0.00885	0.23621		
Plant height at maturity (cm)	-0.06846	0.00349	0.32697		
Total number of tillers per plant	-0.10418	-0.01127	0.04356		
Number of productive tillers per plant	-0.01300	-0.03579	0.05238		
Length of primary panicle length (cm)	-0.06909	0.05654	0.05918		
Number of grains per panicle	-0.08675	0.03647	0.36960		
Length of seeds (mm)	0.14514	0.81323	-0.35903		
Breadth of seeds (mm)	0.01885	-0.38381	-0.42312		
L/B ratio of seeds	-0.09926	0.33285	0.11113		
Weight of hundred seeds (g)	0.12028	-0.01302	0.20638		
Total dry matter production per plant (g)	0.10945	-0.07700	0.41181		
Seed yield per plant (g)	-0.09888	-0.05208	-0.12610		
Harvest index (%)	0.29368	-0.25637	-0.37972		

In the pooled analysis, principal component 1 accounted for 69 per cent of the morphological variation in the rice germplasm collection and was loaded on days 50 per cent flowering and harvest index. Principal component 2 accounted for 11 per cent of the

morphological variation in the rice germplasm collection and was loaded on length of sees. The remaining variables had weak or no discriminatory power. Thus, the must important descriptors were those associated with principal component 1 and 2.

The result amply indicated that the most important descriptors were those associated with principal component 1 and 2. It would be useful in evaluating the potential breeding value of the rice germplasm.

## References

- Adams, M.W. (1995). An estimate of homogeneity crop plants with special reference to genetic variability in dry season *Phaseolus vulgaris*. *Euphytica*, **26:** 665-679.
- Amy, E.L. and M.P. Pritts (1991). Application of principal component analysis to horticultural research. *Hort. Science*, 26(4): 334-338.
- Bharadwaj, C.H., S.C. Tara and D. Subramanyam (2001). Evaluation of different classificatory analysis methods in some rice (*Oryza sativa* L.) collections. *Ind. J. Agric. Sci.*, **71(2):** 123-125.
- Hotelling, H. (1933). Analysis of a complex of statistical variable into principal components. *J. Educ. Psych.*, 24: 417-441.
- Kotaiah, K.C., C. Panduranga, C. Rao, N. Sreevarma, N. Reddi and R.B. Sarma (1986).
  Mahalonobis D<sub>2</sub> and metroglyph analysis in mid-duration genotypes of rice. *Ind. J. Agric. Sci.*, 56(3): 151-156.
- Maji, A.T. and A.A. Shaibu (2012). Application of principal component analysis for rice germplasm characterization and evaluation. *J. Pl.Breeding and Crop Sci.*, 4(6): 87-93.
- Rao, C.R. (1964). The use and interpretation of principal component analysis in Applied Research. Sankhya A, 26: 329-358.
- Raychaudhuri, S., S.M. Stuart and R.R. Altman (2000). Principal components analysis to summarize micro-array experiments. Application to sporulation time series. Pacific Symposium on Bio-computing.
- Smith, L.I. (2002). A tutorial on principal components analysis; http:// www.cs.otago.ac.nz/ cosc453/ student\_tutorials/principal\_components/ pdf.