PHENOTYPIC STABILITY IN ELITE PIGEON PEA (*CAJANUS CAJAN* (L.) MILL.) GENOTYPES USING BIPLOT MODELS

N SANDHYA KISHORE*, P JAGAN MOHAN RAO, A SARITHA, G NEELIMA, S SANDEEP, P MADHUKAR RAO, SRIDHAR CHOUGHAN AND N LINGAIAH

Professor Jayashankar Telangana State Agricultural University, Hyderabad, India

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Abstract

The yield stability of promising pigeon pea genotypes across six locations was assessed using additive main effects and multiplicative interaction and Genotype plus genotype and environment interaction biplot models. The combined analysis of variance indicated that yield was significantly affected by genotypes, environments and genotype and environment interaction. The first two interaction principal components from the AMMI analysis accounted for 68.48% and from GGE accounted for 64.4% of variation due to GxE interaction. Both AMMI and GGE biplots identified G3 and G4 as most stable and G4 with the highest yield across six locations. The “What-won-where” biplot revealed that six environments fell into four sectors i.e., E1 and E2 fell into one mega environment (ME 1), E3 and E6 another (ME 2), E4 in one mega environment (ME 3) and E5 into another (ME 4). Environments E5 and E1 were most representative while E1 could be regarded as least representative environment.

Introduction

Pulses are the important component of food chain and feed system across the globe as they contribute immensely to food and nutritional security in a sustainable manner through the diversification of agricultural production system (Singh et al. 2020). Food legumes supplement nutritionally rich fodder and feed for ruminants. Therefore, they serve as a functional element of many viable cropping and farming systems. Pigeon pea (*Cajanus cajan* (L.) Millspaugh) is commonly known as red gram, tur and arhar. It is the second most important pulse crop after chickpea and occupies an important place under rainfed agriculture in India (Sameer Kumar et al. 2014). It is a crop with a large temporal variation (97-299 days) for grain maturity. Hardy, widely adapted and drought tolerant traits allow its cultivation in a wide range of environments and different cropping systems.

GxE interactions are a challenge to plant breeders because they cause difficulties in selecting genotypes evaluated in diverse environments. When GxE interaction is significant, its cause, nature, and implications must be carefully considered (Kang et al. 1991). AMMI analysis combines ANOVA and principal component analysis (PCA). The interpretation of results obtained from AMMI analysis is performed with a biplot that relates to genotypic means to the first or some of the principal interaction components. GGE biplot analysis enables visual (graphical) presentation of the interaction estimate. GGE analysis enables the identification of genotypes with the highest yields in different environments, comparison of their performances in different environments, identification so called “ideal” genotype, as well as “mega environments”

There are two types of biplots that have been extensively used to visualize G × E interactions, and these are the AMMI (Gauch 1988, Gauch and Zobel 1996, GGE biplots, Yan et al. 2000). However, the application of the two biplots in research has been debated by Gauch et al. (2008). The comparison studies of GGE biplot with AMMI Models validate the equal efficiency of GGE

*Author for correspondence: <kishoregene@gmail.com>.
biplot and AMMI models in understanding G×E interaction (Dias et al. 2003). However, the difference between these models is the omission of environmental component in GGE biplot (Yan et al. 2007, Gauch et al. 2008). The present study aimed to estimate the pattern of genotype × location interaction for few elite medium duration pigeon pea genotypes to take a decision on their potential and adaptability during rainy (kharif) season for different locations of Telangana state and also to do a comparative study of popular multivariate analysis models, i.e., additive main effects and multiplicative interaction model (AMMI) and genotype, genotype × environment interaction (GGE) model.

**Materials and Methods**

The experiments were conducted during the rainy season, 2019, at six research stations viz., Agricultural Research Station, Tandur (E1), Regional Agricultural Research Station, Palem (E2), Agricultural Research Station, Adilabad (E3), Regional Agricultural Research Station, Warangal (E4), Regional Agricultural Research Station, Jagtial (E5) and Agricultural Research Station, Tornala (E6) working under Professor Jayashankar Telangana State Agricultural University, Telangana state, India. The characteristics of the different sites are presented in Table 1. Fig. 1 shows the amount of rainfall received at six different locations over the trial evaluation period.

**Table 1. Parentage details of Pigeon pea genotypes along with environmental conditions.**

| Sl. No. | Code | Genotype | Parentage | Code | Environments | Latitude | Longitude | MSL   |
|--------|------|----------|-----------|------|--------------|----------|-----------|-------|
| 1      | G1   | WRG-366  | ICPL-85063 × ICP-7035 | E1   | Tandur       | 17° 15'N  | 77° 35'E  | 553.18m |
| 2      | G2   | WRG-374  | WRG-65 × ICP-7035     | E2   | Palem        | 16°35'N  | 78°10'E  | 642.00 m |
| 3      | G3   | WRG-65   | WRG-13 × ICPL-87051   | E3   | Adilabad     | 19°39'N  | 78°32'E  | 268.80 m |
| 4      | G4   | WRG-369  | WRGE-14 × BSMR-736    | E4   | Warangal     | 18° 03'N | 79° 22'E | 270.00 m |
| 5      | G5   | WRG-368  | WRG-53 × BSNR-736     | E5   | Jagtial      | 18°49'N  | 78° 56'E | 243.40 m |
| 6      | G6   | WRG-367  | WRG-65 × BDN-2004-2   | E6   | Tornala      | 18° 06'N | 78° 44'E | 483.80 m |
| 7      | G7   | IBTDRG-8 | ICPL-87119 × ICPW-29   |      |              |          |           |       |
| 8      | G8   | ICPL-87119 | C11 × ICPL 6         |      |              |          |           |       |

Fig. 1. Rainfall received during trial evaluation period over six environments.
The genotypes evaluated in the present investigation were developed by pedigree method of breeding at Regional Agricultural Research Station, Warangal and Agricultural Research Station, Tandur under Professor Jayashankar Telangana State Agricultural University, Telangana State, India. The selected homozygous genotypes were evaluated in station yield trials from 2016-2018 for grain yield and also screened for traits like Fusarium wilt and Sterility mosaic Disease (SMD) resistance. Based on the superior performance, a total of eight entries viz., WRG-366 (G1), WRG-374 (G2), WRG-65 (G3), WRG-369 (G4), WRG-368 (G5), WRG-367 (G6), IBTDRG-8 (G7) and ICPL-87119 (G8) were included in the multilocation testing.

Tillage at each site initiated with summer ploughings with discs (Depth adjust to 25 cm) followed by harrowing. The seeds were treated with *Rhizobium* culture. Sowing was carried out after receiving field capacity of rainfall (60mm). Field trials were carried out using RCBD with three replicates. Each genotype was planted in four rows with a length of 4 m and spaced 120 cm apart between the rows and 10 cm between plants. Each replication containing eight genotypes and 2 meters between the replications. At 15 days after emergence, the plots were thinned to one plant per hill. Cultural practices such as weed, fertiliser and plant protection operations were executed. Drainage channels were laid around the trials for collecting the run-off. Comprehensive pest management methods, including the use of pheromone traps, bird perches were used. Standard crop management practices recommended by the state agricultural university were followed in raising a healthy crop across the locations. Data were recorded for grain yield on plot basis.

The combined analysis of variance was preceded to study the genotype x environment interaction of the genotypes across all environments. The AMMI model which combines standard analysis of variance with PC analysis was used to study the interactions (Zobel et al. 1988).

The AMMI model for G genotypes and E environments can be written as -

\[ Y_{ij} = \mu + g_i + e_j + \sum_{n=1}^{N} \alpha_i \lambda_n + \lambda_{jn} + \theta_{ij} \]

Where,

- \( Y_{ij} \) is the mean yield of \( i^{th} \) genotype in the \( j^{th} \) environment;
- \( \mu \) is the general mean;
- \( g_i \) is the \( i^{th} \) genotypic effect;
- \( e_j \) is the \( j^{th} \) location effect;
- \( \lambda_n \) is the eigen value of the principal component axis \( n \);
- \( \alpha_i \) and \( \lambda_{jn} \) are the \( i^{th} \) genotype, \( j^{th} \) environment principle component analysis (PCA) scores for the PCA axis \( n \);
- \( \theta_{ij} \) is the residual.

\( n \) is the number of PCA axis retained in the model.

A windows based software GEA-R9 genotype x environment analysis with R for (Windows) version 4.1 developed by CIMMYT, Mexico was used for AMMI and GGE biplot analysis (Pacheco et al. 2015). The number of PCA axis to be retained is determined by testing the mean square of each axis with the estimate of residual through F-statistic (Gollob 1968, Gauch 1988). GGE biplot analysis was carried out by using the AMMI model R-packages 1.5 PBTools 1.4 version IRRI (PBTools 2014). The AMMI biplot is developed by placing both genotype of environment mean value on X-axis and the respective IPCA axis eigenvector on the Y-axis (Zobelet al. 1988).
Results and Discussion

The mean grain yield over the locations ranged from 1826 kg/ha (WRG-368) to 2033 kg/ha (WRG-369), while, maximum grain yield (2344 kg/ha) recorded at E6 (Tornala) and minimum at 1601 kg/ha at E1 (Tandur) among the different environments (Table 4). Combined analysis of variance of eight medium duration pigeon pea genotypes tested for grain yield over six locations indicated that pigeon pea grain yield was significantly (p<0.01) affected by environments and genotypes × environment interactions (Table 2) indicating the presence of considerable interaction of genotypes with the environments for the trait under study. Only a small portion of (2.23%) the total sum of squares was attributed to genotypic effects. 52.59% of the total sum of squares was ascribed by environmental fluctuations exhibiting that the environments were diverse, with large difference among environmental means causing the most of the variation in yield. This is in harmony with the findings of Panwar et al. (2008) and Mahnaz Rashidi et al. (2013). Genotypes × environment interactions significantly explained 26.02% of the treatment’s variation for grain yield. The combined analysis of variance indicated that there was a strong interaction between genotypes and environments. The trend of mean grain yield (kg ha⁻¹) over environments is graphically represented in Fig. 2.

Table 2. Combined analysis of variance of grain yield for eight medium duration Pigeon pea genotypes evaluated at six environments.

| Source                | d.f | S.S       | M.S       | Explained S.S (%) |
|-----------------------|-----|-----------|-----------|-------------------|
| Varieties             | 7   | 553003.34 | 79000.47  | 2.23              |
| Environments          | 5   | 13010710.24 | 2602142.04 | 52.59            |
| Varieties X Environments | 35  | 6438347.89 | 183952.79 | 26.02            |
| Error                 | 96  | 4737467.24 | 49348.61  |                   |
| Total                 | 143 | 24739528.72|           |                   |

![Mean grain yields over six environments](image)

Fig. 2. Yield performance of eight Pigeon pea genotypes over six environments.

To investigate the main effects and interactions across different environments, AMMI 1 and AMMI 2 biplots were constructed for yield (Table 3). AMMI 1 biplot of main effects (genotype and environments) are shown along the abscissa and the ordinate represents the first IPCA. The interpretation of biplot assay is that if main effects have IPCA score close to zero, it indicates negligible interaction effects and when a genotype and an environment have the same sign on the IPCA axis, their interaction is positive, if different, their interaction is negative. The IPCAI versus IPCAII biplob (AMMI 2 biplot) explains the magnitude of interaction of each genotype and
environment. Earlier reports confirmed that in most of the cases the maximum genotype and environment interaction could be explained through using the first two PCAs (Fikere et al. 2014 and Biswas et al. 2019). Therefore, IPCA1 and IPCA2 were used for construction of AMMI1 and AMMI2 biplots. The genotype and environments that are farthest from the origin being more responsive and fit the worst genotypes and environments that fall into the same sector interact positively and negatively if they fall into opposite sectors (Anandan et al. 2009). The IPCA scores of genotypes in the AMMI analysis are an indication of stability over environments (Gauch and Zobel 1996 and Martin and Alberts 2004).

Table 3. AMMI analysis of variance for grain yield of eight medium duration Pigeon pea genotypes evaluated at six environments.

| Source     | S.S     | Percent | Percent ac. | DF | M.S     | F   |
|------------|---------|---------|-------------|----|---------|-----|
| ENV        | 13010059.15 | 65.04   | 65.04       | 5  | 2602011.82 | 52.72|
| GEN        | 552723.60 | 2.76    | 67.81       | 7  | 78960.51  | 1.59 |
| ENV*GEN    | 6438361.68 | 32.18   | 100         | 35 | 183953.19 | 3.72 |
| PC1        | 2788049.15 | 43.30   | 43.30       | 11 | 253459.01 | 5.51 |
| PC2        | 1621502.26 | 25.18   | 68.48       | 9  | 180166.91 | 3.91 |
| PC3        | 1257139.94 | 19.52   | 88.01       | 7  | 179591.42 | 3.90 |
| PC4        | 515516.38  | 8.00    | 96.02       | 5  | 103103.27 | 2.24 |
| PC5        | 256153.93  | 3.97    | 100         | 3  | 85384.64  | 1.85 |
| PC6        | 0        | 0       | 100         | 1  | 0        | 0   |
| Residuals  | 4737856  | 0       | 0           | 96 | 49352.66  |     |

According to the AMMI model, the genotypes which are characterized by means greater than striking mean and the IPCA score nearly zero are considered as generally adaptable to all environments. According to Fig. 3, the genotypes G4,G1 and yielded above the overall genotype average, whereas the genotype G5 was the poor yielder. The Genotypes G4, G2 and G5 exhibited high grain yield than mean yield with specific adaptability for the environments E5 (Jagtial) and E3 (Adilabad). As the genotypes and environments of this adaptive group have the same sign on the IPCA axis. Genotype G1 revealed specific adaptation for the environment E5 with high grain yield more than mean yield and positive interaction. Genotypes G3 and G5 exhibited specific adaptability for the environment E4 with grain yield less than mean. Varieties with high mean yield and large PCA scores were considered as explicitly adapted to specific environments (Abdi and Williams 2010 and Askari et al. 2017). For the AMMI 2 biplot, IPCA 2 scores considered in interpreting genotype × environment interaction as suggested by Gauch and Zobel (1996). A biplot is generated using genotypic and environmental scores IPCA2, Purchase et al. (2000) pointed out that the closer the genotypes score to the center of the biplot, the more stable the genotypes are (Fig. 4). Allocation of genotype points in the AMMI 2 biplot revealed that the genotypes G3 and G4 dotted close to the origin, indicating least interaction of these genotypes with environments. The remaining genotypes scattered away from the origin in the biplot indicating that the genotypes were more sensitive to environmental interactive forces. The same trend has been observed by Mahnaz Rashidi et al. (2013). The genotypes G5 and G2 had positive interaction with environments E6 and E3. The genotypes G1 and G8 displayed positive interaction with the environment E5. The genotype G7 indicated specific adaptability and positive interaction with the environments E1 and E2.
Table 4. Mean yields (kg/ha) of the pigeon pea genotypes across six environments.

| Genotype & Environment Code | Genotype & Environments names | Mean yield (kg/ha) | AMMI IPCA1 | IPCA2 | GGE IPCA1 | IPCA2 |
|-----------------------------|--------------------------------|--------------------|------------|-------|-----------|-------|
| G1                          | WRG-366                        | 1980               | -7.22      | 13.60 | -242.16   | 379.87|
| G2                          | WRG-374                        | 1978               | 5.58       | -11.07| 161.46    | -165.63|
| G3                          | WRG-65                         | 1876               | 5.35       | -1.59 | 189.12    | -190.35|
| G4                          | WRG-369                        | 2033               | 3.70       | 5.43  | 89.30     | 217.07|
| G5                          | WRG-368                        | 1826               | 16.21      | -4.31 | 525.60    | -154.84|
| G6                          | WRG-367                        | 1946               | 8.22       | 7.65  | 245.74    | 242.19|
| G7                          | IBTDRG-8                       | 1918               | -15.96     | -16.49| -481.37   | -459.55|
| G8                          | ICPL-87119                     | 1894               | -15.89     | 6.78  | -487.69   | 131.23|
| E1                          | Tandur                         | 1601               | -22.82     | -7.32 | -0.770    | -0.08 |
| E2                          | Palen                          | 1605               | -1.67      | -6.98 | -0.08     | -0.152|
| E3                          | Adilabad                       | 2236               | 12.88      | -10.30| 0.38      | -0.21 |
| E4                          | Warangal                       | 1725               | 13.96      | 9.24  | 0.39      | 0.59  |
| E5                          | Jagtial                        | 2077               | -7.36      | 20.39 | -0.27     | 0.73  |
| E6                          | Tornala                        | 2344               | 5.020      | -5.01 | 0.14      | -0.18 |

Fig. 3. Biplot of the first interaction principal component axis (IPCA1) versus mean yields.
The What-won-where GGE biplot (Fig. 5) revealed the information about which genotype is top performer in the respective environment and ideal for which particular environment. The polygon is drawn on genotypes that lie farthest from the biplot origin in such way that all other genotypes are contained within the polygon. The “what-won-where” biplot revealed that six environments fell into four sectors i.e., E1 and E2 fell into one mega environment (ME 1), E3 and E6 another mega environment (ME 2), E4 in one mega environment (ME 3), and E5 into another mega environment (ME 4). As E1 and E2 fell under single ME 1 and E3 and E6 under ME 2, from each MES, any one of these environments is sufficient to study the performance of genotypes. Genotype 7 was the winner in the mega environment 1, Genotype G5 was the winner in mega environment 2, genotype G6 was winner in mega environment 3 and genotype G1 was winning genotype in mega environment 4. These genotypes performed better than other genotypes in the respective mega environments.

The average environment axis (AEA) is the line passing through the average environment and the biplot origin. A test environment showing a smaller angle with the AEA is more representative than other test environments (Yan and Rajcan 2002). Test environments that are consistently non-discriminative provide little information on the genotypes, which is inappropriate as test environments (Kang-Bo-Shim et al. 2015). Discriminativeness vs Representativeness GGE biplot for grain yield revealed that E5 had the least cosine angle with average environmental axis (AEA) (Fig. 6) and could be declared as the least discriminative environment, whereas E4 was observed to be the most discriminative environment as it made the largest cosine angle with AEA followed by E1. Environment E6 and E2 were the test environments that had both discriminating and representative, hence could serve as good test environment for selecting generally adapted genotypes. Discriminative but non-representative test environment E4 was useful for rejecting unstable genotypes if the target environment is a single mega environment. E5 was most representative among all environments, while E1 was least representative of all environments.
Fig. 5. Polygon view of the GGE biplot of grain yield of eight genotypes over six locations.

Fig. 6. GGE biplot based on environment focused for comparing environments with ideal environment.

The line that passes through the biplot origin and the average environment with single arrow is the average environment axis (AEA). Projections of genotype markers to the AEA showed the mean yield of genotypes. Greater projection onto AEA ordinate, regardless of the direction means greater instability (Kang-Bo-Shim et al. 2015). The genotypes G3, G2 and G4 showed shorter projections and were relatively stable over environments (Fig. 7). Whereas, G7 showed greater projection onto AEA ordinate indicating the greater instability. The genotype that showed good performance with stability including G4 since it ranked first in yielding and short projection of the genotype marker lines.
Fig. 7. Biplot of stability and mean performance of eight genotypes across average environments.

The genotypes G4, G1 and G2 yielded above the overall genotypes average, whereas the genotype G5 was the poor yielder. The Genotypes G4, G2 and G5 exhibited high grain yield than mean yield with specific adaptability for the environments E5 and E3. Environments E2 and E6 could be regarded as more stable sites and the environment E4 as the most discriminating environment while E1 was the most representative environment. Hence, the genotype G4 showing most stable across six locations with highest yield can be given to farmers field for further evaluation before its release for general cultivation.

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