



Research Article

Genetic Studies for Detection of Most Diverse and High Yielding Genotypes among Chickpea (*Cicer arietinum* L.) Germplasm

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Abstract | Genetic variation occurring naturally in germplasm is highly valuable resource of alleles that can be deployed for genetic improvement of a species. Screening of available genetic stock for detection of most diverse and high yielding genotypes is a pre requisite for a successful crop breeding program. For this purpose, a research experiment comprising of sixty-eight elite chickpea germplasm genotypes along with two commercial varieties were sown under tri-replicate randomized complete block design during the winter season of 2020-21. D² statistics, principle component analysis and cluster analysis were employed to detect the most genetically variable and high yielding chickpea genotypes. D² statistics extracted higher values for standard deviation and coefficient of variation indicating that the studied genotypes possess considerable amount of genetic variation in performance of studied different traits. Principle component analysis distinguished the traits into eight components. Results revealed that PC1 and 2 extracted >1 Eigen values explaining that these components have major contribution in genetic variability. Cluster analysis distributed the genotypes into four distinguished clusters. Agglomerative dendrogram of genotypes was constructed by Ward's method. On the basis of Euclidean distance it was observed that members of cluster 3 (G.P-110, G.P-111, G.P-115, G.P-116, G.P-121, G.P-125, G.P-138, G.P-139, G.P-150, G.P-151, G.P-164, G.P-165) and cluster 4 (G.P-112, G.P-126, G.P-140, G.P-152, G.P-166) were most diverse. Results also showed that the genotypes of cluster 3 were high yielding and genetically diverse as well. Therefore, they may be preferred while making selections for chickpea genetic improvement program.

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Keywords | Chickpea, Germplasm, Cluster analysis, PCA, Variability

Introduction

Chickpea (*Cicer arietinum* L.), is third most grown pulse legume crop across the globe. It is an important food legume crop of tropical and subtropical climates of world (Gautam *et al.*, 2021). Domestication of chickpea by humans dates back to 10,000 years ago (Abbo *et al.*, 2007). About 6000

years ago, its cultivation spread from Fertile Crescent of southeastern Anatolia, Turkey (Rajeev *et al.*, 2019). Besides its nutritional value, chickpea improves the soil fertility through symbiotic nitrogen fixation. Recently chickpea is being grown in more than 50 countries of the world (Tsehaye *et al.*, 2020). Pakistan ranks 3rd in chickpea producing countries but far below than world's average chickpea productivity per

unit area (Shah *et al.*, 2020). Its yield potential has been seriously hampered due to narrow genetic base and several biotic and abiotic stresses. Narrow genetic base and lack of high yielding commercial cultivars are the most important factors for declining chickpea yields per unit area. Most of the commercial varieties of chickpea are vulnerable to climatic changes and have limited adaptability to diversified environments (Agrawal *et al.*, 2018). Existence of variability among individuals guarantees the survival of species in nature and pre requisite for a successful crop genetic advancement program (Govindaraj *et al.*, 2015; Singh *et al.*, 2021).

Genetic variation among individuals of a species is highly useful for breeders to identify and select the most desirable genotypes for the breeding programs (Bulti and Haji, 2019; Tsehaye *et al.*, 2020). Estimation of genetic variation and its exploitation in hybridization programs is highly valuable for a successful crop breeding program (Rafiq *et al.*, 2020). Among germplasm, each genotype has a specific genetic make-up differing from other genotypes therefore, provide more options to researchers for selection of desirable genotypes (Farshadfar and Farshadfar, 2008; Johnson *et al.*, 2015).

Exploration of genetic variability of germplasm and assessment of patterns of genetic diversity assists the breeders to identify and select the most appropriate parental types from a mixed population. (Govindaraj *et al.*, 2015). D² statistics, principal component and cluster analysis have been found very important series statistical methods for variability assessment of large scale crop data and widely utilized by several crop breeders (Malik *et al.*, 2014; Chen *et al.*, 2017; Singh *et al.*, 2021).

Limited scientific progress has been made so far in Pakistan for screening and characterization of available chickpea germplasm (Sani *et al.*, 2018; Shah *et al.*, 2020). Therefore, the current study was planned for detection of the most diverse and high yielding genotypes among chickpea germplasm.

Materials and Methods

The present study for screening of genetic variability among sixty-eight elite chickpea germplasm lines along with two standard varieties was carried out at 71.165 °E and 32.920 °N (Gram Breeding Research,

Station, Kallurkot, Pakistan) during 2020-21 crop season. Experimental material was laid down under tri-replicate RCBD design, keeping 4 rows each 30 cm apart and 10 cm distance between plants. Dibbler was used for sowing of entries by dropping two seeds in each hole and thinning was done after two weeks to ensure proper plant population. Repeated chemical spray of Emamectin Benzoate @ 800 ml per hectare was done to protect the crop from *Helicoverpa armigera* attack. Hoeing, weeding and irrigation were performed as per requirement of crop.

Data were recorded for wilt percentage, days required for 50 percent flowering, secondary branches, height of plants, number of pods, maturity days, hundred grain weight and yield kg ha⁻¹. D² statistics was employed following the method proposed by Mahalanobis (1928, 1936). For Cluster analysis and Principal component analysis STAR version 2.0.1 (Statistical Tool for Agricultural Research) was employed.

Table 1: Mean performance of different traits of chickpea germplasm.

Traits	Range	Mean (μ)	Standard Deviation (σ)	CV values
Wilt percentage	1-13	5.79	8.28	13.3
Days to 50 % flowering	94-114	105.3	9.24	16.5
Secondary branches	7-19	12	12.5	18.7
Plant height (cm)	50-89	62.5	9.48	19.5
Number of pods	36-106	66.9	20.56	26.5
Maturity days	141-165	151.2	8.86	15.4
100 Grains weight (g)	22.2-28.4	24.9	2.62	5.8
Yield kg ha ⁻¹	850-1650	1204	230.98	28.3

Results and Discussion

D² statistics revealed higher values for standard deviation, coefficient of variation and wide ranges values indicating that the studied genotypes varied significantly in performance of different traits (Table 1). From the results it was obvious that sufficient amount of variation exists among the studied genotypes. Our results were in line to (Syed *et al.*, 2012; Malik *et al.*, 2014) who also reported wide dispersion of data for range and coefficient of variation.

Principal component analysis (PCA) distributed the data into eight different components. Results showed

Table 2: Eigen values of variables in Principal Components.

	PC1	PC2	PC3	PC4	PC5	PC6	PC7	PC8
Eigen values	2.2997	1.0161	0.8333	0.7039	0.4973	0.3818	0.2368	0.1990
Percentage of variance	66.11	12.91	8.68	6.19	3.09	1.82	0.70	0.50
Cumulative percentage	66.11	79.02	87.70	93.89	96.98	98.80	99.5	100

Table 3: Extraction of variance in PCA.

Variables	PC1	PC2	PC3	PC4	PC5	PC6	PC7	PC8
Wilt %	-0.3770	0.1047	-0.3957	0.1686	-0.6367	0.1593	-0.2565	0.4066
DF	-0.3279	0.0125	0.2002	-0.8996	-0.0046	0.0424	-0.1326	0.1532
SB	0.3984	0.0346	-0.1116	-0.2601	-0.6543	0.1851	0.3992	-0.3881
PH	0.1042	-0.9408	-0.1464	-0.684	-0.1163	-0.2272	-0.0738	0.0852
NPP	0.4234	-0.0089	0.0476	-0.0514	0.1295	0.3208	0.3471	0.7586
MD	-0.2845	-0.2176	0.8111	0.2931	0.2270	0.2711	0.0467	0.0280
GW	0.3814	0.2306	0.3470	-0.003	-0.2866	-0.6627	-0.3175	0.2425
YLD	0.4184	-0.0437	0.0350	-0.0369	0.0505	0.5209	-0.7269	-0.1351

DF: Days to flowering; **SB:** Secondary branches; **PH:** Plant height; **NPP:** Number of pods plant⁻¹; **MD:** Maturity days; **GW:** 100 grains weight; **YLD:** Yield kg ha⁻¹.

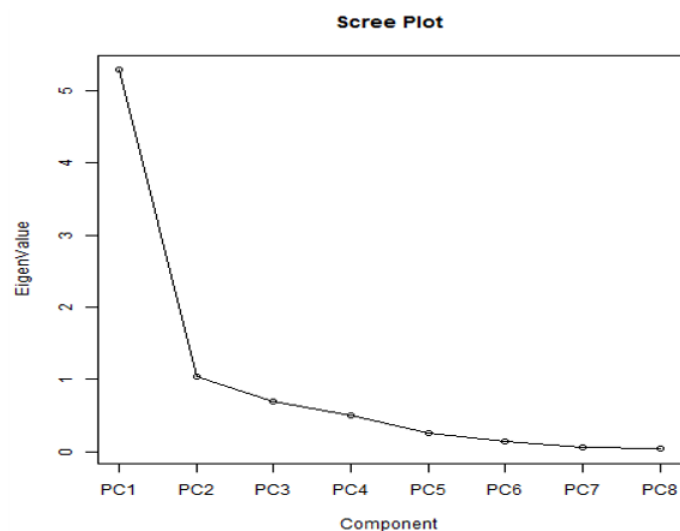


Figure 1: Scree plot showing Eigen values of principle components.

that PC1 and 2 extracted > 1 Eigen values (2.299 and 1.0161 respectively) with a cumulative share of 79 percent in variation (Table 2). Highest percentage of variance was shared by PC1 (66.11 %) followed by PC2 (12.91 %), PC3 (8.68%) and PC4 (6.19%). A Scree plot was constructed for illustration of Eigen values of all components (Figure 1). Our results agree to Arora et al. (2018) and Mahmood et al. (2018). PCs presenting > 1 Eigen values have significant contribution towards variability the PCs having <1 values have non-considerable contributions towards genetic variability (Abdi and Williams, 2010).

Data also revealed that in PC1 number of pods, yield kg ha⁻¹, secondary branches and 100 grain weight ex-

tracted higher positive loadings indicating that maximum amounts of genetic variability was shared by these traits while wilt percentage, days to flowering and days taken to maturity contributed negatively (Table 3). Similarly, in PC2, grain weight, secondary branches wilt percentage and days to flowering showed positive loadings while negative contributions were exhibited by plant height, maturity days, yield kg ha⁻¹ and number of pods plant⁻¹. Our findings were in line to Lal et al. (2001) and Bisht et al. (2005).

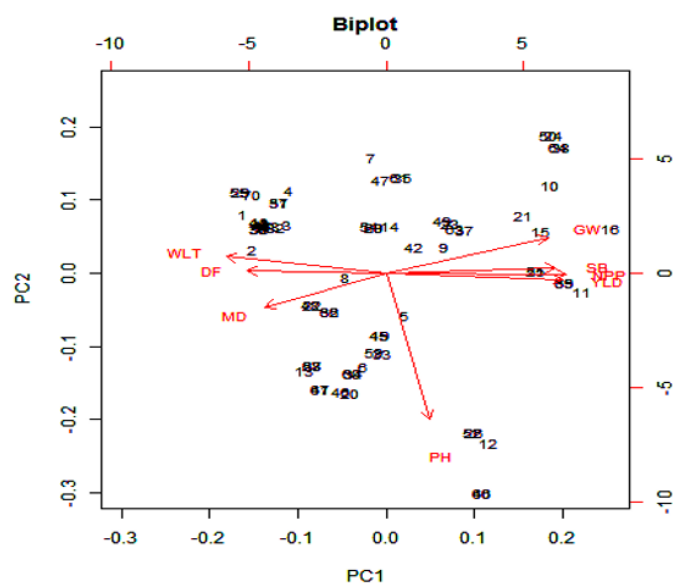


Figure 2: Biplot among PC1 and 2.

Biplot was also constructed among PC1 and 2 (Figure 2). Biplot illustrated that the vectors for number

Table 4: Mean and range values of various traits in different clusters.

Traits	Cluster-1		Cluster-2		Cluster-3		Cluster-4	
	Range	Mean	Range	Mean	Range	Mean	Range	Mean
Wilt	7-13	10.4	3-8	5.5	1-2	1.62	3-4	3.4
DF	103-113	108.8	101-114	106.7	94-104	97.6	101-104	103
SB	7-11	8.5	9-18	11.5	16-19	17.1	13-16	14.6
PH	52-60	54.9	50-74	63.7	52-70	59.9	83-89	86
NPP	36-48	41.2	52-78	64.7	96-106	100.5	80-84	82.4
MD	150-156	153.4	144-165	153.1	141-152	144.9	142-149	146.9
GW	22.2-24.6	23.6	22.8-26.8	24.5	26.4-28.4	27.3	25.4-25.9	25.7
YLD	850-1060	918	970-1380	1184	1460-1650	1554	1340-1460	1398

DF: Days to flowering; **SB:** Secondary branches; **PH:** Plant height; **NPP:** Number of pods plant⁻¹; **MD:** Maturity days; **GW:** 100 grains weight; **YLD:** Yield kg ha⁻¹.

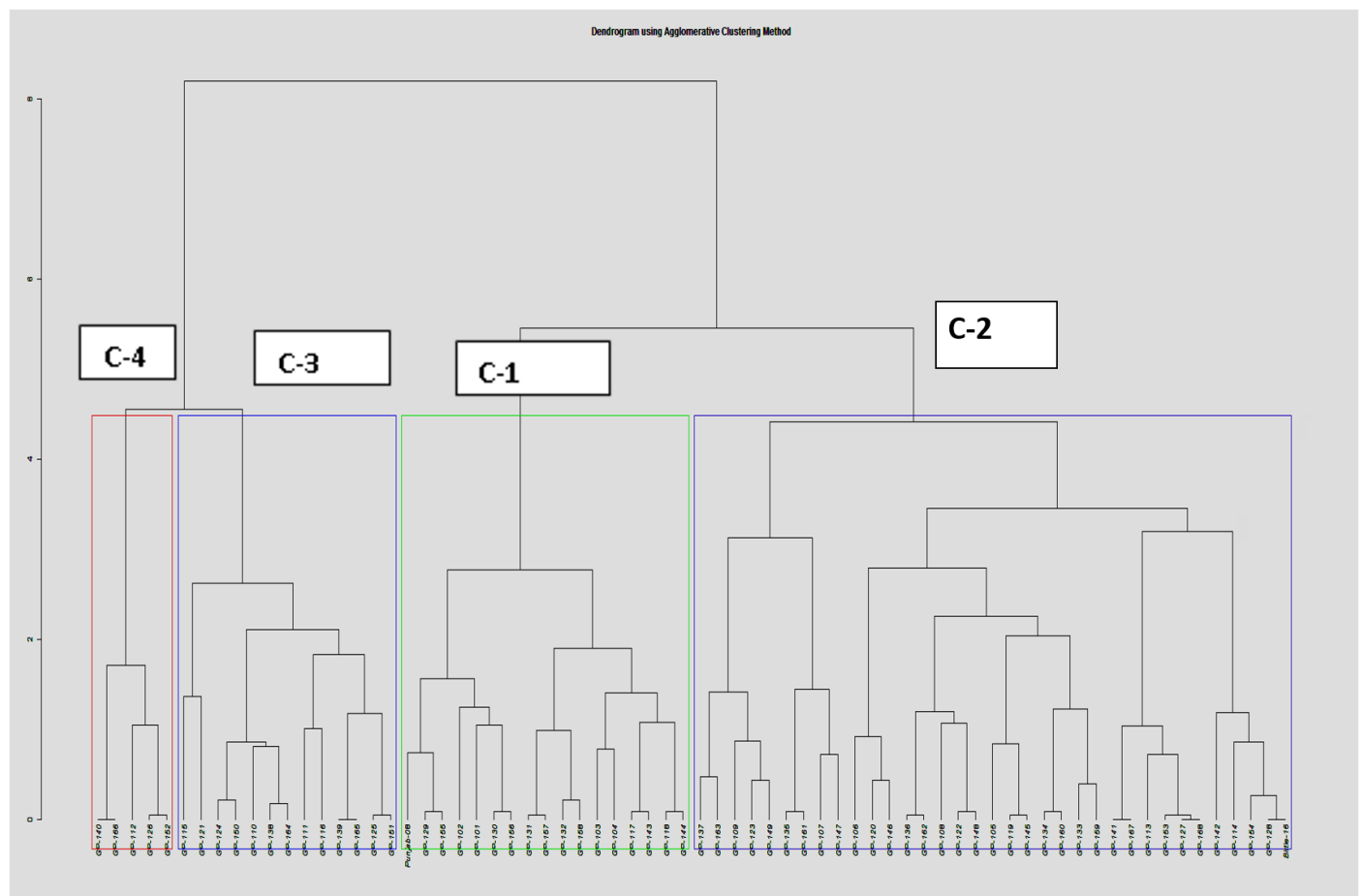


Figure 3: Ward's Dendrogram of Agglomerative clustering.

of pods, yield kg ha⁻¹, secondary branches and 100 grain weight were superimposed on plots indicating that these traits are most influential variable in construction of PC1. In PC2 grain weight, secondary branches wilt percentage and days to flowering exhibited positive contributions while all other traits contributed negatively. Our findings agree to the previous finding of Zubair *et al.* (2017), Sharifi *et al.* (2018) and Singh *et al.* (2021).

Cluster analysis distributed the genotypes into four distinguished clusters. Data revealed that the genotypes with maximum yield mean (1554 kg ha⁻¹) and high values of other yield contributory traits *i.e.* maximum grain weight (27.3 g), number of pods (100), secondary branches (17) were grouped under cluster-3 followed by cluster-4 (Table 4). Likewise, maximum wilt (10.4 %), days to flowering (109) were recorded in cluster-1. Our findings agree to Upadhyaya *et al.* (2007), Pavan *et al.* (2017) and Singh *et al.* (2021).

Table 5: Cluster Membership of different genotypes.

Clusters	No of members	Membership
Cluster-1	17	G.P-101, G.P-102, G.P-103, G.P-104, G.P-117, G.P-118, G.P-129, G.P-130, G.P-131, G.P-132, G.P-143, G.P, 144, G.P-155, G.P-156, G.P-157, G.P-158, Punjab-2008.
Cluster-2	35	G.P-105, G.P-106, G.P-107, G.P-108, G.P-109, G.P-113, G.P-114, G.P-119, G.P-120, G.P-122, G.P-123, G.P-27, G.P-128, G.P-133, G.P-134, G.P-135, G.P-136, G.P-137, G.P-141, G.P-142, G.P-147, G.P-148, G.P-149, G.P-153, G.P-154, G.P-159, G.P-160, G.P-161, G.P-162, G.P-163, G.P-167, G.P-168, Bittle-2016
Cluster-3	13	G.P-110, G.P-111, G.P-115, G.P-116, G.P-121, G.P-125, G.P-138, G.P-139, G.P-150, G.P-151, G.P-164, G.P-165
Cluster-4	5	G.P-112, G.P-126, G.P-140, G.P-152, G.P-166

Ward's dendrogram of agglomerative clustering was also constructed on the basis of Euclidean distance (Figure 3). Germplasm entries were distributed among four distinguished clusters. Data also showed that 17 genotypes were grouped in cluster-1, 35 were gathered under cluster-2, 13 genotypes in cluster-3 and 5 in cluster-4 (Table 5). Dendrogram showed that maximum Euclidean distance was observed in cluster-3 and 4. Therefore the members of the cluster 3 (G.P-110, G.P-111, G.P-115, G.P-116, G.P-121, G.P-125, G.P-138, G.P-139, G.P-150, G.P-151, G.P-164, G.P-165) and cluster 4 (G.P-112, G.P-126, G.P-140, G.P-152, G.P-166) are most diverse in performance of different studied traits. On the other hand it was also found that the members of cluster 3 were diverse and high yielding as well, therefore they may be favored while making selections for chickpea genetic improvement program. Our results relate to the previous findings of Talebi and Rokhzadi (2013); Zubair *et al.* (2017) and Mahmood *et al.* (2018).

Conclusions and Recommendations

From the study following conclusions may be derived;

1. Wide dispersion of data for standard deviation, range and coefficient of variation were evident that the germplasm entries varied significantly in performance of different traits.
2. PCA revealed that sufficient amount of genetic

diversity exists among studied traits. Results also revealed that the vectors for number of pods, yield kg ha⁻¹, secondary branches and 100 grain weight extracted higher positive loadings and were superimposed on plots demonstrating that these traits are most influential in creation of genetic variability.

3. Cluster analysis distributed the genotypes into four distinguished clusters. Results indicate that the members of cluster-3 (G.P-110, G.P-111, G.P-115, G.P-116, G.P-121, G.P-125, G.P-138, G.P-139, G.P-150, G.P-151, G.P-164, and G.P-165) are more diverse in performance of different traits and high yielding as well. Therefore, these genotypes may be favored while making selections for chickpea genetic advancement program.

Novelty Statement

Detection of most diverse and better performing genetic material from gene pool of land races is novel study and will be helpful for researchers and chickpea breeders.

Author's Contribution

Muhammad Tariq Mahmood: Conceived idea and wrote manuscript.

Muhammad Akhtar: Technical inputs, supervised and overall management of research study.

Kaiser Latif Cheema: Wrote results and discussion section.

Abdul Ghaffar: Did statistical analysis and interpretation of results.

Muhammad Jahanzaib Khalid: Recorded research data

Imtiaz Ali: Checked Plagiarism and necessary corrections.

Zeshan Ali: Did revision, citation and references section of the article.

Conflict of interest

The authors have declared no conflict of interest.

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