Genotype by environment interaction and yield stability in sugarcane

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Interacción genotipo x ambiente y estabilidad del rendimiento en caña de azúcar

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ABSTRACT

Keywords:
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Hybrid

Genotype by environment interaction (GEI) reduces the association between phenotype and genotype which result in relative ranking and stability differences of genotypes across environment. The objectives of this research were (i) to select sugarcane genotypes of high yield and stable (ii) to study the interrelationships among various parametric and no parametric stability statistics. Seventeen experimental genotypes and three check cultivars of sugarcane were evaluated at seven environment using randomized completely block design. Methodologies based on analysis parametric (Regression-bi-S^2d_i, Shukla variance, Ecovalence-W, Coefficient of variation-CV, index of Lin and Binns-PI and AMMI value) and non-parametric statistics (Nassar and Huehn- $Si^{(1)}$, $Si^{(2)}$, $Si^{(3)}$, $Si^{(6)}$, Kang-RS, Fox-TOP, and Thennarasu- $NPi^{(1)}$, $NPi^{(2)}$, $NPi^{(3)}$, $NPi^{(4)}$) were used for Ton of Pol per hectare (TPH). Genotypes and environment showed high significant difference ($P<0.01$) while GEI was significant ($P<0.05$). The parametric stability analysis identified the genotypes V99-236 and V00-50 as the most stable and high TPH. With non-parametric statistics were identified the genotypes V00-50, V99-236 and V98-120 as most stable. The analysis distinguished two groups of statistics using biplot: the first group (G1) formed by PI, CV, ASV, TOP, $Si^{(3)}$, $Si^{(6)}$, $NPi^{(2)}$, $NPi^{(3)}$ and $NPi^{(4)}$ statistics were located under the concept of dynamic stability since they are associated with TPH. The other group (G2), formed by Shukla, W, S^2d_i, bi, RS, $Si^{(2)}$, $Si^{(1)}$, $NPi^{(1)}$ statistics, fell within the static concept. Finally, genotypes V99-236 and V00-50 can be recommended as the most stable genotype in terms of both stability andTPH.

RESUMEN

Palabras clave:
Interacción G x A
Estabilidad fenotípica
Correlación de rango
Saccharum spp.
Híbrido

La interacción genotipo por ambiente (GEI) reduce la asociación entre el fenotipo y el genotipo lo cual genera cambios en el orden y en la estabilidad de genotipos a través de ambientes. Los objetivos de esta investigación fueron: (i) seleccionar genotipos de caña de azúcar de alto rendimiento y estables (ii) evaluar las interrelaciones entre diversos métodos de estabilidad paramétrica y no paramétrica. Diecisiete genotipos experimentales y tres cultivares testigos de caña de azúcar fueron evaluados en siete ambientes utilizando un diseño de bloques completamente aleatorizado. Metodologías basadas en el análisis estadístico paramétrico (Regression-bi-S^2di, varianza de Shukla, Ecovalence-W, Coeficiente de variación-CV, índice de Lin y Binns-PI y AMMI) y no paramétrico (Nassar and Huehn- $Si^{(1)}$, $Si^{(2)}$, $Si^{(3)}$, $Si^{(6)}$, Kang-RS, Fox-TOP, and Thennarasu- $NPi^{(1)}$, $NPi^{(2)}$, $NPi^{(3)}$, $NPi^{(4)}$) fueron usadas para evaluar el rendimiento en toneladas de Pol por Hectárea (TPH). Los genotipos y el ambiente mostraron diferencias estadísticamente significativa ($P<0,01$), mientras que la GEI fue significativo ($P<0.05$). Los estadísticos de estabilidad paramétricas identificaron los genotipos V99-236 y V00-50 como los más estables y de alto TPH y los no paramétricos distinguieron a los genotipos V00-50, V99-236 y V98-120. El biplot identifico dos grupos de estadísticos: El primer grupo formado por los estadísticos PI, CV, ASV, TOP, $Si^{(3)}$, $Si^{(6)}$, $NPi^{(2)}$, $NPi^{(3)}$, y $NPi^{(4)}$) que se situaron bajo el concepto de estabilidad dinámica, ya que están asociados con TPH. El otro grupo (G2), formado por los estadísticos Shukla, W, S^2di, bi, RS, $Si^{(2)}$, $Si^{(1)}$, $NPi^{(1)}$ caen dentro del concepto estabilidad estática. Finalmente, los genotipos V99-236 y V00-50 pueden ser recomendados como los más estables y de alto TPH.

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Genotype by environment interaction (GEI) is a relevant consideration for plant breeders. Sugarcane breeders usually assess a group of genotypes through environments prior to the release of a new crop for farmers' production (Rea and De Sousa-Vieira, 2002). The GEI causes the best genotype to change with the environment, and makes the selection process difficult for a particular region (Farshadfar *et al.,* 2012). Most breeders have used the term 'stability' to describe a genotype which exhibits a relatively constant yield, independent of environmental conditions. This concept of stability is in accordance with the concept of homeostasis extensively used in quantitative genetics (Becker and Leon, 1988) and may be regarded as a 'biological' or 'static' concept of stability. A genotype showing an unchanging performance in all environments does not certainly respond to improved growing conditions with increased yield. Agronomists, hence, would prefer an 'agronomic' or 'dynamic' concept of stability in which it is not indispensable that the genotypic response to environmental conditions should be identical for all genotypes (Becker and Leon, 1988).

Diverse methods have been proposed to estimate GEI or yield stability. These methods can be separated into two main groups including parametric (univariate and multivariate) and non-parametric approaches based on different strategies (Dehghani *et al.,* 2008). Univariate parametric methods have been used, such as those of Eberhart and Russell (1969), Shukla (1972) and Francis and Kannenberg (1978). These methods demand few computation and their parameters are easy to interpret biologically.

Diverse nonparametric methods have been used to describe and explain the responses of GEI (Nassar and Huehn, 1987; Kang, 1988; Fox *et al.,* 1990; Thennarasu, 1995). In this approach, no suppositions about the observations are required and there is less susceptible to measurement errors or to outliers (Huehn, 1990; Balalić *et al.*, 2011; Temesgen *et al.,* 2015; Scampin *et al.,* 2000; Rea *et al.,* 2015).

Multivariate methods such as the additive main effects and multiplicative interaction (AMMI) model has been proposed as effective methodology to predict adaptation and stability of cultivars (Guerra *et al.,* 2009). Purchase

et al. (2000) generated the AMMI stability value (ASV) based on the AMMI model and using principal component scores (axes 1 and 2) for each genotype.

Mixed model methods also can be used to estimate GEI effects when analyzing multilocation yield trial data. If locations are random representatives of environments in the target region, the mixed model procedure stipulates best linear unbiased predictors (BLUP) of random effects. The BLUP of random effects is suitable for detecting location specific effects and estimating genotype by location interaction effects (Coutiño-Estrada and Vidal-Martínez, 2003).

The objectives of this research were (i) to select sugarcane cultivars of high cane yield (TPH) and stable through different environments in Venezuela (ii) to study the interrelationships among various parametric and no parametric phenotypic stability statistics.

MATERIAL AND METHODS

Seventeen experimental genotypes and three commercial check cultivars of sugarcane were evaluated at seven environment using randomized completely block design with three replications in the Central-Western region of Venezuela. The experimental sugarcane genotypes were: V91-1, V91-2, V91-6, V91-8, V91-15, V98-62, V98-86, V98-120, V99-117, V99-190, V99-203, V99-208, V99-213, V99-217, V99-236, V99-245 and V00-50. The check cultivars evaluated were B80-408, C323-68, and CP74-2005. All materials were evaluated at seven locations (Carora and Montaña Verde in Lara State; The Majaguas, Farm Ivonne and Farm Castillera in Portuguesa State; Santa Lucia and FUNDACANA in Yaracuy State), each with three crop-years (plant crop, first and second ratoon) during 2008-2010. Some environmental conditions of the seven experimental sites of Venezuela can be seen in Table 1. The attribute evaluated was cane yield expressed in tons of pol per hectare (TPH). The plots were three rows, with 1.5 m between rows and 10 m long. Plots were managed conventionally and followed the established local practices. All three rows were harvested for measuring cane yield (TCH). The cane was incinerated and then chopped by hand. A 10-stalk sample was randomly taken from each plot and weighed. The samples were milled and the crusher juice was analysed for sucrose content (Pol % cane). Tons of pol

per hectare (TPH=t pol ha^{-1}) was estimated as related to tons of cane per hectare (TCH) and Pol % cane by the coming formula: TPH = (TCH x Pol % cane)/100 (Guerra *et al.,* 2009).

Table 1. Principal soil and precipitation characteristics of the appraised locations

Location	Soil	Precipitation (mm)	pH
Quebrada arriba (A)	Clay loam	1101	7.7
Santa Lucia (B)	Silty clay loam	700	8.0
FUNDACAÑA (C)	Silt loam	1111	8.1
Montaña verde (D)	Loam	1048	7.3
Las Majaguas (E)	Clay loam	1500	7.0
Finca Ivone (F)	Clay loam	1500	7.0
Finca Castillera (G)	Clay loam	1500	7.0

Analysis of variance. The analysis of variance was executed contemplating the genotype and environment effects as fixed, according to the mathematics model: Yijk = µ + B/Ejk + Gi + Ej + GEij + εijk, where Yijk represents the ith genotype within the jth environment and the kth block, µ is the general mean, B/Ejk corresponds to the block within the jth environment in the kth block, Gi is the effect of the ith genotype, Ej is the effect of the jth environment, GEij is the effect of interaction of the ith genotype with the jth environment, and εijk is the effect of experimental error. The GEI was divided as stated by the additive main effect and multiplicative interaction (AMMI) models (Crossa, 1990). Data were combined over locations and analysed as combined series of RCB´s with repeated measures (crop-year) using InfoStat software (Balzarini *et al.*, 2008).

Stability analysis. Stability of the 20 genotypes for TPH was calculated by using the coefficient of regression (*b*), mean squared deviations from regression (S²dᵢ), ecovalence stability index (*W*)**,)** Shukla's stability variance (Shukla), Linn and Binn Index (Pi), the coefficient of variation (CV) and AMMI stability value (ASV). Several nonparametric stability statistics suggested by Nassar and Huehn (1987); Kang (1988); Fox *et al.* (1990) and Thennarasu (1995) were estimated.

The statistics based on yield ranks of genotypes (Nassar and Huehn, 1987) in each environment are expressed as follows: $S_i^{(1)}$ calculates the average of the absolute differences in the orders of a genotype in all environments, $S_i^{(2)}$ is the variance between the ranks in all environments, $S_i^{(3)}$ and $S_i^{(6)}$ are the sum of the absolute deviation and sum of squares of ranks for each genotype relative to the average of the ranks, respectively (Rea *et al.*, 2015) .

$$S_i^{(1)} = 2\sum_{j}^{n-1} \sum_{j'=j+1}^{n} \left[r_{ij} - r_{ij'} \right] / \left[n(n-1) \right]$$

$$S_i^{(2)} = 2\sum_{i=1}^{n} (r_{ij} - \overline{r_{i.}})^2 / (n-1)$$

$$S_i^{(3)} = \sum_{j=1}^{n} (r_{ij} - \overline{r_{i.}})^2 / \overline{r_{i.}}$$

$$S_i^{(6)} = \sum_{i=1}^{n} \left| r_{ij} - \overline{r_{i.}} \right|^2 / \overline{r_{i.}}$$

Where r_{ij} is the rank of the ith genotype in the jth environment, and $\overline{r_{i.}}$ is the mean rank across all environments for the ith genotype.

Thennarasu's (1995) nonparametric stability analysis considers adjusted ranks of genotypes within each test environment.

$$NP_i^{(1)} = 1/n \sum_{j=1}^{n} \left| r_{ij}^* - M_{di}^* \right|$$

$$NP_i^{(2)} = 1/n \left(\sum_{j=1}^{n} \left| r_{ij}^* - M_{di}^* \right| / M_{di} \right)$$

$$NP_i^{(3)} = \frac{\sqrt{\sum \left(r_{ij}^* - \overline{r}_{i.}^* \right) / n}}{\overline{r}_{i.}}$$

$$NP_i^{(4)} = 2/n(n-1) \left[\sum_{j=1}^{n-1} \sum_{(j'=j+1)}^{n} \left| r_{ij}^* - r_{ij'}^* \right| / \overline{r}_{i.} \right]$$

The adjusted rank r_{ij}^* is calculated on the basis of the adjusted phenotype values $\left(X_{ij}^* = X_{ij} - \overline{X}_{i.} \right)$ where $\overline{X}_{i.}$ is the mean performance of the ith genotype. The ranks, obtained from these adjusted values (X_{ij}^*), depend only on GEI and error effects; r_{ij}^* is the rank X_{ij}^*, $\overline{r}_{i.}^*$ and M_{di}^* are the mean and median ranks for adjusted values, while $\overline{r}_{i.}$ and M_{di} are the same parameters computed from the original (unadjusted) values.

Fox et al. (1990) suggested a non-parametric superiority procedure for general adaptability using stratified ranking of cultivars. A genotype that appeared mostly in the top third (high *TOP*-value) was judged a widely adapted cultivar. Kang's (1988) rank-sum (*RS*) is another non-parametric stability procedure where both yield and Shukla's (1972) stability variance were used as selection criteria. In this method, both the highest yielding genotype and the genotype with the lowest stability variance are ranked

1 and the genotype with the lowest RS value is judged the most desirable (Farshadfar et al., 2012; Rea et al., 2015).

Additionally, Spearman correlation coefficients among stability parameters and principal component analyses (PCA) based on the correlation matrix were executed to achieve an understanding of the association among stability parameter. All of the analyses were effectuated using InfoStat software (Balzarini et al., 2008) and Agricolae program (De Mendiburu, 2015) which was originated in R software (CRAN).

RESULTS AND DISCUSSION
Analysis of GE interaction
Plant breeders inevitably face GEI once testing varieties across a number of environments. Reckoning on the magnitude of the interactions or the differential genotypic responses to environment, the varietal rankings can dissent greatly across environment (Dehghani et al., 2008; Crossa, 1990). Bartlett´s homogeneity test evidenced that the mean squares of individual environments were unvarying and so the combine analysis of variance was performed. The combined analysis of variance (ANOVA) for TPH is shown in Table 2. Genotypes and environment showed high significant difference (*P*<0.01) while GEI was significant (*P*<0.05) indicating rank difference in genotypes response at different environments and the need for extension of stability analysis.

Table 2. AMMI analysis for TPH (t pol ha⁻¹) of twenty sugarcane genotypes in seven environments.

Sources of variation	GL	SSAMMI	MS-AMMI	% SS
E/Rep	14	470.82	33.63	
Genotype (G)	19	5620.01	295.79**	53.07
Environment (E)	6	2700.16	450.03**	25.50
G x E (GE)	114	2270.12	19.91*	21.43
PCA 1	24	234.025	9.75	31.00
PCA 2	22	146.864	6.68	19.42

*,** significant at 5% and 1% probability level by F test, respectively

The mean TPH across environments over three years (Table 3) showed considerable changes in ranks among the genotypes, reflective the presence of high GE interactions (Rea et al., 2015). These results show that the heterogeneousness of the environments and

genetic variability of the genotypes becomes manifest. The environment mean yield ranged from 13.12 TPH in Santa Lucia to 17.56 TPH in Montaña Verde indicating differences among test environments. The highest yield 23.27 TPH was obtained from genotype V98-120 at

Montaña Verde, while the lowest was 8.26 TPH from genotype V91-6 at Santa Lucia. The results of the non-parametric and parametric statistics and their ranks are presented in Table 4 and 5, respectively.

Table 3. Mean yield (TPH) of twenty sugarcane genotypes tested across crops in seven environments.

Genotype/ Location	Quebrada Arriba	Santa Lucia	FUNDACANA	Montaña Verde	The Majaguas	Ivonne	Castillera
B80-408	13.90	11.93	16.30	17.92	16.44	15.00	14.17
C323-68	14.99	13.40	17.11	20.27	18.19	15.66	13.98
CP74-2005	14.97	12.11	14.00	15.1	15.54	12.84	12.34
V00-50	18.29	14.46	17.08	20.62	19.28	15.69	16.6
V91-1	15.19	13.10	14.61	18.68	12.31	13.79	10.84
V91-15	10.09	11.62	11.85	16.01	13.85	11.06	12.29
V91-2	12.03	10.43	13.38	13.12	12.71	9.88	8.63
V91-6	15.36	8.26	14.18	15.17	13.86	12.17	11.56
V91-8	14,24	12.19	11.09	14.03	12.84	12.04	11.62
V98-120	18.31	15.46	16.57	23.27	17.16	14.01	16.06
V98-62	19.64	14.87	17.43	21.38	19.01	13.73	13.44
V98-86	14.39	12.68	11.64	14.36	14.64	14.71	12.02
V99-117	13.04	11.53	15.96	14.16	14.09	11.25	12.44
V99-190	18.48	16.81	14.67	18.06	14.22	13.73	15.27
V99-203	17.52	16.20	14.23	16.06	15.07	14.27	12.49
V99-208	19.78	15.29	14.93	19.93	18.80	14.65	16.76
V99-213	15.70	14.60	18.03	21.49	18.12	15.08	16.47
V99-217	12.80	10.93	10.20	14.27	10.99	13.51	11.53
V99-236	20.78	16.68	20.44	20.08	18.78	15.72	16.24
V99-245	12.46	9.78	14.46	17.17	9.55	11.20	12.89
Mean Location	15.60	13.12	14.91	17.56	15.27	13.50	13.38

Univariate stability. In Table 4 is presented the mean yield values (TPH) and stability parameters. Finlay and Wilkinson (1963) and Eberhart and Russell (1966) considered genotypes with high mean yield, coefficients of regression equivalent to unity (bi= 1) and deviation from regression proximate zero (S^2d_i=0) to be stable. According to these parameters only V99-236 and V00-50 genotypes meet these conditions. Francis and Kannenberg (1978) defined a stable genotype as one that provides high yield and constant performance across locations. In conformity with this definition, V99-236, V99-190, V99-213, V98-62 and V98-120 are considered stable since that presented low CV and high TPH. Rea and De Sousa-Vieira (2002) used this method and concluded that the CV could be used to identify genotypes on a group basis rather than individually; however, the method can also be used in a plant-breeding context. In the method of Lin and Binns (1988) the best genotype is considered to be the one with the greatest performance and the lowest value of Pi. Here, we found that genotypes V99-236, V00-50, V98-120, V98-62, and V99-208 showed lower Pi values, indicating greater adaptability to these environments. Wricke (1962) recommended using ecovalance (W) as a stability parameter. Genotypes with the smallest ecovalance (W) values are contemplated stable. The W was lowest for genotypes CP74-2005, V00-50 and V99-236. Shukla (1972) defined a stability variance value, which considered a genotype with relatively large variance to have low stability. By Shukla´s definition,

genotypes CP74-2005, V00-50 and V99-236 ranked the most highly stables. Considering all methods of univariate stability can conclude that genotypes V99-236 and V00-50 were the most stable and high yields.

Multivariate stability. The AMMI model does not make specification for a quantitative stability measure, and as such a measure is important in order to quantify and sort genotypes in terms of yield stability. Hence, AMMI stability value **(**ASV) suggested by Purchase *et al* (2000) was applied to quantify and rank genotypes according to their yield stability. The genotype with the lowest ASV value is taken as most stable. Consequently, genotypes V99-236, CP74-2005 and B80-408 were the most stable (Table 4).

Table 4. Mean yield values (TPH) and stability parameters of twenty sugarcane genotypes across seven environments.

Genotypes	TPH	PCA1	PCA2	ASV	*bi*	S^2d_i	CV	W	Shukla	Pi
V91-1	14.07	0.02	0.94	6.95	1.29	1.57	32.60	40.58	7.15	187.54
V91-2	11.45	-0.09	0.39	6.26	0.91	1.23	27.33	22.91	3.87	413.82
V91-6	12.94	-0.36	0.49	3.64	1.30	1.57	30.57	40.93	7.21	276.37
V91-8	12.58	0.88	-0.25	5.12	0.53	0.92	28.18	29.56	5.11	318.45
V91-15	12.39	-0.44	-0.93	4.98	0.80	1.65	37.73	42.89	7.57	339.22
V98-62	17.07	-0.25	1.37	4.84	1.87	1.06	28.71	51.23	9.12	32.76
V98-86	13.49	0.90	-0.71	3.94	0.36	1.32	25.69	44.92	7.95	259.88
V98-120	17.26	-0.45	0.41	5.18	1.75	1.19	28.81	46.37	8.23	30.85
V99-117	13.21	-0.39	0.06	3.49	0.64	1.44	30.47	36.59	6.41	256.55
V99-190	15.89	1.09	0.13	4.67	0.62	1.78	26.85	54.16	9.66	97.54
V99-203	15.12	1.26	0.21	4.65	0.49	1.59	32.07	49.31	8.76	139.80
V99-208	17.16	0.49	0.45	5.32	1.15	1.57	29.50	38.01	6.67	46.26
V99-213	17.07	-0.97	-0.31	5.51	1.29	1.35	27.52	30.83	5.34	41.55
V99-217	12.03	0.73	-1.07	5.82	0.45	1.46	29.71	45.64	8.08	385.28
V99-236	18.39	0.03	-0.01	1.34	1.09	1.40	26.30	22.51	3.80	10.73
V99-245	12.50	-0.77	-0.49	5.41	1.16	2.17	38.61	71.51	12.87	321.91
V00-50	17.43	-0.31	0.14	5.23	1.25	0.84	31.72	13.44	2.12	30.06
B80-408	15.10	-0.74	-0.60	2.99	0.96	1.40	28.58	29.47	5.09	132.13
C323-68	16.23	-0.86	-0.29	4.15	1.32	1.38	28.54	33.41	5.82	74.26
CP74-2005	13.84	0.22	0.10	1.90	0.77	0.79	30.21	11.85	1.83	206.08

Genotypes	Pi	TOP	RS	Si(1)	Si(2)	Si(3)	Si(6)	NPi(1)	NPi(2)	NPi(3)	NPi(4)
V91-1	187.54	0	22	0.71	44.14	13.02	2.85	4.86	0.44	0.71	0.08
V91-2	413.82	0	24	0.29	58.62	6.47	3.88	4.00	2.00	1.90	0.12
V91-6	276.37	0	27	0.24	29.81	10.73	2.50	4.00	0.57	0.80	0.04
V91-8	318.45	0	19	0.24	30.57	7.60	2.40	4.57	0.91	1.02	0.05
V91-15	339.22	0	31	0.43	38.00	8.41	3.12	4.57	0.76	1.17	0.09
V98-62	32.76	4	23	0.00	24.81	5.07	1.33	6.29	0.35	0.45	0.00
V98-86	259.88	0	27	0.52	44.90	11.65	2.60	5.14	0.64	0.76	0.06
V98-120	30.85	1	19	0.38	32.48	2.38	0.78	3.86	0.24	0.34	0.02
V99-117	256.55	0	23	0.33	34.00	10.94	2.94	4.43	0.89	0.80	0.05
V99-190	97.54	1	27	0.48	47.95	7.59	1.79	5.71	0.48	0.48	0.04
V99-203	139.80	1	26	0.43	45.95	5.98	1.59	5.29	0.48	0.53	0.04
V99-208	46.26	3	14	0.67	52.14	3.27	1.06	5.86	0.37	0.41	0.04
V99-213	41.55	3	13	0.48	32.90	2.17	0.89	4.71	0.28	0.32	0.03
V99-217	385.28	0	34	0.14	40.57	7.92	2.77	5.00	1.25	1.59	0.04
V99-236	10.73	4	7	0.14	30.81	1.28	0.64	4.14	0.22	0.28	0.01
V99-245	321.91	0	37	0.43	58.95	19.33	4.77	6.43	2.14	1.28	0.08
V00-50	30.06	3	4	0.24	22.62	2.24	0.82	3.43	0.20	0.26	0.01
B80-408	132.13	0	15	0.29	34.33	7.41	1.85	4.14	0.32	0.47	0.02
C323-68	74.26	1	15	0.48	44.24	3.55	1.20	5.43	0.34	0.42	0.03
CP74-2005	206.08	0	13	0.24	17.62	3.25	1.25	3.00	0.38	0.49	0.03

Non parametric stability. The results of 10 nonparametric stability measures and genotypes mean yield are resumed in Table 4. The Si[1] and Si[2] statistics are based on ranks of genotypes across locations and they give proportional weight to each environment. Genotypes with less modification in rank are considered to be more stable. According to both Si [1] and Si [2] V99-236 and V00-50 had the smallest changes in ranks and are thus considered as the most stable genotypes. Two other non-parametric statistics, Si [3] and Si [6], integrate yield and stability based on yield orders of genotypes in each environment (Nassar and Huehn, 1987). The lowest value for each of these statistics reveals maximal stability for a certain genotype. Genotypes V99-236, V00-50, V98-120, V99-208, V99-213 had the lowest Si [3] and Si [6] values hence, these genotypes were identified as the most stable genotypes.

Results for Thennarasu´s (1995) non-parametric stability statistics, estimate from ranks of adjusted yield means, are showed in Table 4. The ranks of genotypes based on these statistics are presented in Table 5. According to NPi[1], genotypes CP74-2005, V00-50, V98-120 were stable in analogy with the other genotypes. Genotypes V00-50, V99-236, V98-120, V99-236 had the lowest value of NPi[2] and were judged stable. NPi[3] and NPi[4] also recognized genotypes V00-50, V99-236 and V98-120 how the most stable genotypes and high mean yield. Kang´s (1988) non-parametric stability statistic (RS) applies both yield and stability variance (Shukla, 1972) with the genotype having the lowest rank-sum being the most promising. In this case, the genotypes V00-50, V99-236, V99-213, and CP74-2005 had the lowest values, and were stable genotypes with high yield in comparison with other genotypes. Non-parametric (TOP) superiority measure (Fox *et al.*, 1990) identified genotypes V99-236, V98-62, V00-50 and V99-208 presented mainly in the top third, thus, these genotypes were stable. Similarly, Segherloo *et al.* (2008) found a highly significant association between mean yield and Fox-rank.

Table 5. Ranks of twenty genotypes across environments using stability statistics

Genotypes	TPH	ASV	bi	S^2d_i	CV	W	Shukla	Pi	TOP	RS	Si[1]	Si[2]	Si[3]	Si[6]	NPi[1]	NPi[2]	NPi[3]	NPi[4]
V91-1	11	20	10	15	18	11	11	11	8	10	20	13	19	16	12	10	12	17
V91-2	20	19	3	6	4	4	4	20	8	13	8	19	10	19	4	19	20	20
V91-6	15	5	11	16	15	12	12	15	8	15	4	4	16	13	4	13	14	9
V91-8	16	12	15	3	6	6	6	16	8	8	4	5	13	12	9	17	16	14
V91-15	18	11	6	18	19	13	13	18	8	18	12	11	15	18	9	15	17	19
V98-62	5	4	20	4	9	18	18	4	1	11	1	3	8	8	19	7	7	1
V98-86	13	6	18	7	1	14	14	14	8	15	18	15	18	14	14	14	13	16
V98-120	3	13	19	5	10	16	16	3	5	8	11	7	4	2	3	3	4	4
V99-117	14	10	13	12	14	9	9	13	8	11	10	9	17	17	8	16	14	14
V99-190	8	9	14	19	3	19	19	8	5	15	15	17	12	10	17	11	9	9
V99-203	9	8	16	17	17	17	17	10	8	14	12	16	9	9	15	11	11	9
V99-208	4	15	4	14	11	10	10	6	3	5	19	18	6	5	18	8	5	9
V99-213	6	17	9	8	5	7	7	5	8	3	15	8	2	4	11	4	3	6
V99-217	19	18	17	13	12	15	15	19	8	19	2	12	14	15	13	18	19	9
V99-236	1	1	1	10	2	3	3	1	1	2	2	6	1	1	6	2	2	2
V99-245	17	16	5	20	20	20	20	17	8	20	12	20	20	20	20	20	18	17
V00-50	2	14	8	2	16	2	2	2	3	1	4	2	3	3	2	1	1	2
B80-408	10	3	2	11	8	5	5	9	8	6	8	10	11	11	6	5	8	4
C323-68	7	7	12	9	7	8	8	7	5	6	15	14	7	6	16	6	6	6
CP74-2005	12	2	7	1	13	1	1	12	8	3	4	1	5	7	1	9	10	6

Correlation between mean yield (TPH) and stability statistics

The Spearman's rank correlations between each combination of stability measures were estimated (Table 6) and demonstrated a high positive significant rank correlation between TPH and PI, TOP, NPi[2], NPi[3], and NPi[4]. The parameters NPi[2], NPi[3] and NPi[4] were associated positive with Si[6] and Si[3] with Si[6], RS, and NPi[2]. Therefore, only one of these parameters would be adequate to select stable genotypes in a breeding

program (Mohammadi *et al.*, 2007). Complete correlation was found between Wricke's and Shukla's statistics. Kang *et al.* (1987) indicated that Wricke´s ecovalance (W) and stability variance (Shukla) were equal; because Shukla´s stability variance is a linear combination of the ecovalence so for ranking purposes these methods are equivalent (Bujak *et al.*, 2014). To better comprehend the interrelationships among the parametric and non-parametric statistics, principal component analysis, based on the correlation matrix of rank (Table 5) was used. The first and second principal components of the rank correlation accounted for 48.10% and 18.80% of the variation, respectively, making a total of 66.90% of the original variance among the stability parameters (Figure 1). Similar results have been reported from other studies in faba bean and field pea (Flores *et al.*, 1998), durum
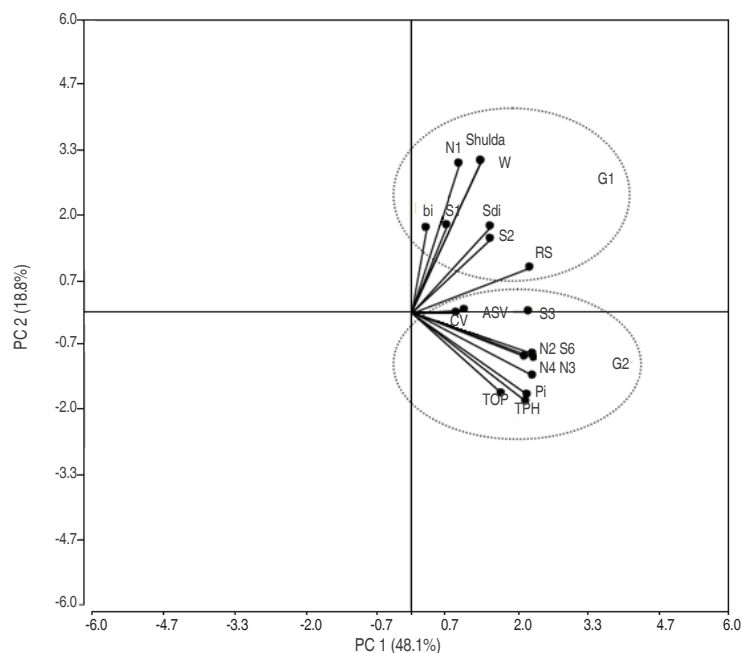

Figure 1. Biplot of IPC1 and IPC2 of the rank correlation matrix of the 17 stability parameters with mean yield (TPH).

Table 6. Spearman correlation of stability parameters and TPH in twenty sugarcane genotypes at seven environments.

Parameters	TPH	ASV	bi	Sdi	CV	W	Shukla	Pi	TOP	RS	$S_i^{(1)}$	$S_i^{(2)}$	$S_i^{(3)}$	$S_i^{(6)}$	$NP_i^{(1)}$	$NP_i^{(2)}$	$NP_i^{(3)}$
ASV	0.28																
bi	-0.02	-0.02															
Sdi	0.27	0.16	-0.15														
CV	0.24	0.26	-0.07	0.39													
W	0.13	0.14	0.57	0.58	0.23												
Shukla	0.13	0.14	0.57	0.58	0.23	1**											
Pi	0.99**	0.3	-0.02	0.3	0.24	0.15	0.15										
TOP	0.8**	0.24	-0.02	0.23	0.27	0.004	0.004	0.79**									
RS	0.71	0.2	0.34	0.62	0.26	0.76	0.76	0.72	0.45								
$S_i^{(1)}$	-0.06	0.35	-0.03	0.42	0.03	0.26	0.26	-0.004	0.16	0.1							
$S_i^{(2)}$	0.3	0.43	-0.14	0.62	-0.03	0.43	0.43	0.36	0.19	0.5	0.65						
$S_i^{(3)}$	0.74	0.21	0.17	0.53	0.35	0.47	0.47	0.74	0.61	0.79**	0.25	0.4					
$S_i^{(6)}$	0.9	0.33	-0.03	0.45	0.33	0.3	0.3	0.89**	0.7	0.78	0.12	0.46	0.89**				
$NP_i^{(1)}$	0.002	0.14	0.26	0.5	-0.003	0.68	0.68	0.03	-0.14	0.42	0.42	0.6	0.32	0.19			
$NP_i^{(2)}$	0.92**	0.33	0.11	0.38	0.21	0.35	0.35	0.93**	0.64	0.8**	0.03	0.47	0.78	0.89**	0.26		
$NP_i^{(3)}$	0.97**	0.3	0.07	0.34	0.26	0.27	0.27	0.98**	0.72	0.79**	-0.05	0.38	0.79**	0.93**	0.11	0.96**	
$NP_i^{(4)}$	0.79**	0.5	-0.09	0.46	0.24	0.22	0.22	0.83**	0.66	0.64	0.44	0.61	0.75	0.85**	0.19	0.84**	0.82**

wheat (Kilic *et al.*, 2010), barley (Mut *et al.*, 2010) and sugarcane (Rea *et al.*, 2015). Results of the biplot of the first two principal components based on the rank correlation matrix were more or less consistent with the Spearman rank correlation coefficients (Figure 1, Table 6). The stability parameters were separated into two stability concepts: Group 1, parameters that corresponded with the dynamic/agronomic stability concept which were associated to mean yield (TPH) such as PI, TOP, $NPi^{(3)}$, $Si^{(6)}$, $NPi^{(2)}$, $NPi^{(4)}$, $Si^{(3)}$, CV and ASV. Group 2, the remaining stability parameters corresponding with the static/ biological stability concept were assigned to RS, $Si^{(2)}$, $Si^{(1)}$, (S^2di), b*i*, Shukla, W, $NPi^{(1)}$ (Figure 1). These parameters, which identify genotypes with high phenotypic stability without due consideration of mean yield, may not be appropriate, as both breeders and farmers incline toward to select genotypes of good yield with perform regularity across environments (Mohammadi and Amri, 2008; Farshadfar *et al.*, 2012; Khalili and Pour-Aboughadareh, 2016).

CONCLUSIONS

Various stability statistics were used in this research for quantifying genotype stability in relation to yield. Both yield and stability should be examined simultaneously to deal the effect of GEI and to accomplish genotype selection more accurate and refined. Several methods have been considered to analyze phenotypic stability although some of them have their limitations and there is no superior method to be recommended in all circumstances. Besides, some methodologies are optional while others are complementary, being able to be used combined. It is also recommendable to use the parametric and non-parametric stability measures jointly since results obtained from the two groups of stability measures can complement each other. Also, to capitalize on the GEI and to select breeding materials adapted to favourable and unfavourable growing conditions, selection of specific cultivars adapted to specific environments appears to be necessary. Finally, both parametric and non-parametric estimates of stability indicated either V99-936 or V00-50 as the genotype most stable and high yield.

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