

# Association Among Traits by Correlations and Path in Maize Genotype Selection

## ABSTRACT

In the **State** of Espírito Santo, family farmers have grown a number of maize varieties for decades, consisting of open-pollinated populations with valuable importance for the livelihood of smallholder farmers. As such, this study intended to analyze the cause and effect **of** associations between agronomic traits for increasing yield in maize populations cultivated in Espírito Santo. A total of 16 accessions of field maize from conservation works in eight different municipalities were evaluated, in a randomized **complete** block design with three replicates, totaling 48 experimental units. Genetic parameters, genotypic, phenotypic, and environmental correlation were estimated, and a path analysis was conducted. In general, the genotypes present genetic variability, showing a significant difference between the genotypes for all the traits analyzed by the F test ( $P \leq 0.05$ ). Insertion height of the first cob, plant height, and final plant stand variables presented heritability ( $h^2$ ) higher than 80%, suggesting that superior genotypes can be achieved. The highest estimate of phenotypic correlation ( $r_p$ ) was found between the insertion height of the first cob with plant height 0.85 and total number of cobs with number of cobs per plant 0.85. For the estimates of genotypic correlation ( $r_g$ ), the highest was of 0.88 between the insertion height of the first cob and the plant height; number of grains per row and number of cobs per plant; total number of cobs and final plant stand. **The final plant stand and the mass of one thousand seeds were observed to be traits that could be the determinants to directly increase the grain yield.**

**Keywords:** *Zea mays* L., heritability, genetic parameters, simultaneous selection.

## 1. INTRODUCTION

Maize (*Zea mays* L.) is one of the most grown cereals worldwide, for its high genetic diversity and wide adaptability [1]. Such a fact, associated with its nutritional characteristics, makes maize also one of the cereals of commercial importance in different countries [2]. Brazil is the third largest maize producer, meaning that the estimated national production for the first 2018/19 harvest is 27.4 million tons and 63.7 million tons for the second one, with a mean yield of  $5.2 \text{ t ha}^{-1}$  in the last harvest [3].

In the **State** of Espírito Santo, family farmers have grown a number of maize varieties for decades, consisting of open-pollinated populations with valuable importance for the livelihood of smallholder farmers. Estimated total

25 contribution for the 2019 harvest is 38.0 thousand t, with yield (approximately  
26 2.8 t ha<sup>-1</sup>) being one of the lowest in the country [3].

27 Open-pollinated populations present lower yield than that of cultivars due to  
28 their wide genetic basis [4]; [5]. While such variability does not ensure high  
29 yield in the short term, it represents a source of alleles favorable for genetic  
30 response to adverse biotic and abiotic factors, conferring greater long-term  
31 yield stability to maize populations [6]; [7]; [8]. This is a knowledge that needs  
32 to be considered so that the conservation of germplasm of these maize  
33 populations **could be** increasingly promoted [9].

34 With the purpose of exploring favorable alleles in these populations to obtain  
35 higher yield values, the study of agronomic traits has been of paramount  
36 importance in plant breeding, due to the possibility of identifying variability in  
37 germplasm and, especially, the chance to select superior accessions for  
38 morpho-agronomic traits of interest [10]; [11]. In this regard, the association  
39 of knowledge on the correlations between agronomic traits allows the  
40 breeder to design strategies that ensure a higher probability of obtaining  
41 superior materials [12].

42 Correlation studies between traits can be conducted through simple  
43 correlations, combined with the study of the genetic parameters involved,  
44 given that the correlations between two traits can be of phenotypic,  
45 genotypic, or environmental nature, in which only genotypic correlations  
46 **associated with high heritability** [13]. Simple correlations, despite being  
47 useful, do not allow conclusions to be drawn about cause and effect  
48 relationships between them, i.e. they do not comprise the direct and indirect  
49 effects of traits on a basic variable. As an alternative, by means of a path  
50 analysis, it is possible to analyze these relationships between the variables,  
51 according to the Wright method [14].

52 In the **State** of Espírito Santo, both the Instituto Federal do Espírito Santo  
53 (Ifes) and the Instituto Capixaba de Pesquisa e Extensão Rural (Incaper)  
54 have been working on studies with maize populations grown in the **State**,  
55 gathering germplasm of these populations and working on breeding  
56 strategies so as to increase grain yield in the **State** [15]; [16]. As such,  
57 studying correlations between agronomic traits in these maize populations  
58 will contribute to the optimization of the strategies used by breeders in the  
59 **State**.

60 In this manner, the purpose of this study was to analyze the cause and effect  
61 associations between agronomic traits for increasing yield in populations of  
62 maize grown in Espírito Santo.

63

## 2. MATERIAL AND METHODS

The experiment was conducted in the Instituto Federal de Educação, Ciência e Tecnologia do Espírito Santo campus of Alegre in 2017. According to the Köppen international classification, the climate of the region is of “Cwa” type, i.e. tropical hot humid, with a cold and dry winter and a hot and rainy summer. The mean temperature of the region is 23.1°C, and the mean total precipitation is 1,341 mm [17].

In the 2017/2018 harvest season, 16 field maize genotypes from conservation work were evaluated in eight different municipalities in the State of Espírito Santo (Table 1). All accessions obtained are described as open-pollinated populations and have been cultivated for some years by smallholder farmers in their respective locations.

Table 1 - Accessions (genotypes) of maize evaluated in accordance with agronomic traits, in the municipality of Alegre, Espírito Santos State, Brazil, and respective municipalities where they were conserved.

Acessions	GEBs	Origin
Aliança	-	Muqui-ES
Asa Branca	IFES Itapina	Itapina-Colatina-ES
Caiano	-	Linhares-ES
Caipira	-	Linhares-ES
Celina	-	Celina - Alegre-ES
BRS Cipotânea	IFES Itapina-ES	Itapina-Colatina-ES
BRS Diamantina	IFES Itapina-ES	Itapina-Colatina-ES
Emcapa 201	INCAPER	Viana-ES
ES001	IFES Itapina	Itapina-Colatina-ES
Fortaleza	-	Muqui-ES
Incaper Capixaba 203	INCAPER	Viana-ES
MA008	IFES Itapina-ES	Itapina-Colatina-ES
Palha Roxa	IFES Alegre-ES	Iúna-ES
Palha Roxa	IFES Alegre-ES	Muniz Freire-ES
Palha Roxa	-	Venda Nova do Imigrante-ES
Sertanejo	IFES Itapina-ES	Itapina-Colatina-ES

GEBs= Germplasm banks; IFES= Instituto Federal do Espírito Santo; INCAPER= Instituto Capixaba de Pesquisa, Assistência Técnica e Extensão Rural.

In order to implement the evaluation test of the genotypes in the field, it was opted for the randomized complete block design with sixteen genotypes (treatments) and three replicates, according to the following statistical model:  $X_{ij} = m + t_i + b_j + e_{ij}$  in which:  $m$  = represents the overall mean,  $t_i$  =

87 treatment effect (genotypes),  $b_j$  = block effect,  $e_{ij}$  = effect of the experimental  
88 error. The experiment comprised 48 experimental units, which were  
89 composed of three 4.0 m length lines spaced at 1.00 m between them. The  
90 evaluation of agronomic traits was performed on 16 plants within the central  
91 line of the plot.

92 During sowing 15 seeds per linear meter were uniformly distributed in furrow.  
93 At 21 days after sowing (DAS), thinning was performed to establish a  
94 population of 5 plants per linear meter of furrow, corresponding to a density  
95 of 50,000 plants  $ha^{-1}$  (adapted to Corrêa et al., 2014). The phytosanitary  
96 treatments followed the technical recommendations for the cultivation, [18].  
97 The maize was naturally dried in the plant, without using desiccants, until it  
98 reached the moisture content of 13%. In February 2018, the harvest was  
99 done manually.

100 The agronomic traits evaluated were as follows: I – Plant height (HEI)  
101 measured from the base to the last flag leaf before tasselling; II – Insertion  
102 height of the first cob (IHC) – measured from the base of the stem to the first  
103 cob; III – Total number of cobs (TNC); IV – final plant stand (FPS); V – Plant  
104 cob diameter (PCD); VI – Stripped cob length (SCL) – measured from the  
105 base to the apex of the cobs; VII – Stripped cob diameter (SCD) – measured  
106 in the middle of each cob; VIII – Number of rows of grain per cob (NRC); IX –  
107 Number of grains per row (NGR); X – Mass of one thousand seeds (MTS); XI  
108 – Yield (YIE); XII – Number of cobs per plant (NCP). All the traits under  
109 evaluation are descriptors established by Biodiversity International [19]

110 Aside from the conventional treatments to control the fall armyworm  
111 (*Spodoptera frugiperda*), leaf sprays were carried out using the Dipel WP  
112 biological insecticide, made from *Bacillus thuringiensis*. The quantity used  
113 was of 500g  $ha^{-1}$  and the applications were made with the aid of a manual  
114 knapsack sprayer with a 20 litre capacity. Mechanical weeding was used to  
115 control the weeds at 30 DAS.

116 For statistical analyses, the phenotypic ( $r_p$ ), genotypic ( $r_g$ ), and environmental  
117 ( $r_e$ ) correlation analyses and linear regression analysis were performed. For  
118 correlations, the following expressions were applied: phenotypic correlations:

119 
$$r_F = \frac{MPG_{XY}}{\sqrt{MSG_X MSG_Y}}; \text{ genotypic correlations: } r_G = \frac{(MPG_{XY} - PMR_{XY})/r}{\sqrt{\hat{\Phi}_{g(X)} \hat{\Phi}_{g(Y)}}} = \frac{\hat{\Phi}_{g(XY)}}{\sqrt{\hat{\Phi}_{g(X)} \hat{\Phi}_{g(Y)}}},$$

120 and environmental correlations:  $r_p = \frac{MPG_{XY}}{\sqrt{MSR_X MSR_Y}}$ . in which,  $MPG_x$  = mean  
121 product between genotypes for traits X and Y;  $MPR_{xy}$  = mean product  
122 between residues for traits;  $MSG_x$  = mean square between genotypes for  
123 trait X;  $MSG_y$  = mean square between genotypes for trait Y;  $MSR_x$  = mean  
124 square between residues for trait X;  $MSR_y$  = mean square between residues

for trait Y;  $\hat{\Phi}_{g(XY)}$  = genotypic covariance estimator;  $\hat{\Phi}_{g(X)}$ ,  $\hat{\Phi}_{g(Y)}$  = estimators of quadratic components associated with genotypic variabilities for traits X and Y, respectively.

The path analysis consisted of studying the direct and indirect effects of the above-mentioned explanatory independent variables (X) on grain yield, main dependent variable (Y). Considering Y to be a complex trait, resulting from the combined action of other traits, the following model can be defined:  $Y = \beta_{1X_1} + \beta_{2X_2} + \dots + \beta_{nX_n} + \varepsilon$ , in which:  $X_1, X_2, \dots, X_n$  are the explanatory variables, and Y is the main variable (or dependent variable). The direct and indirect effects of the explanatory variables are estimated on the main variable. Therefore,  $r_{iy} = p_i + \sum_{j \neq i}^n p_{ij} r_{ij}$  in which: correlation between the main variable (Y) and the i-th explanatory variable;  $p_i$ : direct effect of variable i on the main variable; and  $p_j r_{ij}$ : indirect effect of variable i by means of variable j on the main variable.

The significance of the genotypic correlation coefficient and the b1 of the regression were evaluated by the "t" test, and the bootstrap with 5000 simulations for phenotypic and environmental correlations was applied in accordance with [20].

In order to verify the co-linearity between the traits, a multi co-linearity test was conducted, in line with Montgomery and Peck cited by [21]. Subsequently, it was carried out through the split of the simple correlation coefficients into direct and indirect effects provided by the trail analysis. For all analyses, the computational resources from the Genes program were applied [22].

### 3. RESULTS AND DISCUSSION

All evaluated traits acted in accordance with the assumption of normality of error distribution ( $p = .05$ ) by the Lilliefors test and homogeneity of residual variances ( $p = .05$ ) by the Bartlett test. These results have proved that, normally, the mathematical assumptions needed to conduct the analysis of variance and further studies were satisfied [23].

As a general rule, the genotypes presented genetic variability, showing a significant difference for all the traits under analysis, excluding the stem diameter of the plant, by the F test ( $P=.05$ ) (Table 2). In this regard, [24] observed the variability of field maize, super sweet maize and teosinte maize populations, pointing out that the low plant genetic variability leads to lower genetic gain for breeding programs. The experimental coefficients of variation ( $CV_e\%$ ) showed magnitudes ranging from 5.03, for stripped cob diameter, to 25.14%, for plant stem diameter. In turn, the coefficient of

168 genetic variation ( $CV_{gi}\%$ ) varied from 0.0 to 22.51 for the traits plant stem  
169 diameter and yield, respectively. These values are basically explained by the  
170 variability of the genetic material used in the analysis. Research with other  
171 cultures **has shown** that high variability of genotypes **facilitated** the selection  
172 process of superior material easier [25]; [26]; [27]; [28].

173  
174 As reported by [29], heritability values ( $h^2$ ) above 80% and variation index  
175 above unit 1.0 ensure satisfactory selection gains. For the insertion height of  
176 the first cob, plant height, and final plant stand, this situation was verified  
177 suggesting that, for these variables, superior genotypes can be obtained by  
178 means of simple selection methods, such as mass selection (Table 3).

179 The estimates of phenotypic ( $r_p$ ), genotypic ( $r_g$ ), and environmental ( $r_a$ )  
180 correlation are depicted in (Table 2). Hence, among the 55 pairs of  
181 combinations for the 12 variables evaluated, 13 had significant ( $r_p$ ), 12 to 1 or  
182 5% probability by the "t" test, and only one significant to 1% by the bootstrap  
183 method with 5000 simulations. There was a positive and high variation from  
184 0.50 to 0.85 for 12 pairs. ( $r_g$ ) also presented 13 significant pairs, but 12 by  
185 the bootstrap method with 5000 simulations at 1 or 5% probability, and only  
186 one to 5% probability by the "t" test. For this one, there was a positive and  
187 high variation from 0.53 to 0.88 for 12 pairs. For **environmental** ( $r_e$ ), 23  
188 combinations were significant using the bootstrap method with 5000  
189 simulations at 1 or 5% probability, varying from 0.4 to 0.88 positive pairs  
190 (Table 3).

191 The highest estimate of ( $r_p$ ) was noticed between the insertion height of the  
192 first cob, with plant height 0.85, and the total number of cobs with number of  
193 cobs per plant 0.85. The lowest correlations ( $r_p$ ) were between stripped cob  
194 length with stripped cob diameter 0.50 and stripped cob length and mass of  
195 one thousand seeds 0.50, being significant estimates by the "t" test. These  
196 results corroborate the ones seen by [30], which indicate the variable cob  
197 height as the most influential in yield variation and of greater relevance in  
198 indirect selection for yield. Number of rows of grains per cob and mass of  
199 one thousand seeds showed significant and high value, though negative (**-**  
200 **0.55**). The higher the number of rows, the lower the conversion of  
201 photoassimilates into starch [31].

202 Among the estimates of ( $r_g$ ), the highest one was 0.88 between the insertion  
203 height of the first cob and the plant height; number of grains per row and  
204 number of cobs per plant; total number of cobs; and final plant stand. Also  
205 among the positive estimates, the lowest one was found between the  
206 insertion height of the first cob and the final plant stand, with 0.53. There was  
207 only one significant and high pair, although with a negative value (**-0.57**)  
208 between number of rows of grains per cob and mass of one thousand grains.



209 The genotype selection with a larger diameter and mass of one thousand  
210 grains enables obtaining genotypes with a good yield [32]. Moreover, for ( $r_e$ ),  
211 the highest estimate happened between total number of cobs and number of  
212 cobs per plant, with 0.88. Stripped cob length and stripped cob diameter 0.4  
213 was the lowest positive and significant value. There was a significant and  
214 high pair, however with a negative value between the number of rows of  
215 grains per cob and mass of one thousand seeds (-0.46), (Table 3).

216 The insertion height of the first cob indicated negative correlations ( $r_g$ ) with  
217 stripped cob length of (-0.19), stripped cob diameter of (-0.32), and number  
218 of rows of grains per cob of -0.54; and positive correlations with number of  
219 grains per row of 0.28, number of cobs per plant 0.47, mass of one thousand  
220 seeds of 0.09, and yield of 0.29, still not significant in all cases (Table 3). In  
221 accordance with [21] a non-significant or low magnitude correlation  
222 coefficient does not suggest lack of relationship between two variables, but  
223 absence of a linear relationship between them.

Table 2. Estimates of genetic parameters for phytotechnical traits of 16 maize genotypes in the municipality of Alegre, Espírito Santo State, Brazil, 2017

SV	Mean squares												
	DF	IHC	HEI	PCD	SCL	SCD	NRC	NGR	TNC	FPS	NCP	MTS	YIE
Blocks	2	0.01	0.00	0.59	1.49	2.26	0.36	19.20	134548.68	48868.35	0.01	1722.67	829.53
Varieties	15	0.10**	0.15**	13.15ns	2.80*	17.09**	4.06**	28.63*	203435.78**	93182.08**	0.02**	4422.62**	4449.19**
Residue	30	0.00	0.01	13.32	1.37	4.46	0.86	13.44	44870.97	14574.77	0.01	980.06	1075.33
Genetic parameters													
$(\sigma_f^2)$		0.03	0.05	4.38	0.93	5.69	1.35	9.54	67811.92	31060.69	0.00	1474.20	1483.06
$(\sigma_e^2)$		0.00	0.00	4.44	0.45	1.48	0.28	4.48	14956.99	4858.25	0.00	326.68	358.44
$(\sigma_g^2)$		0.03	0.04	0.0	0.47	4.20	1.06	5.06	52854.93	26202.43	0.00	1147.52	1124.62
$(\hat{h}_g^2)$		90.57	87.17	0.0	51.05	73.87	78.74	53.04	77.94	84.35	64.80	77.83	75.83
$(CV_e\%)$		9.30	6.11	25.14	7.50	5.03	7.41	11.42	16.00	7.59	12.35	11.14	22.01
$(CV_{gi}\%)$		16.66	9.20	0.0	4.42	4.88	8.24	7.00	17.37	10.17	9.67	12.05	22.51
$(CV_{gi}/CV_e)$		1.78	1.50	0.0	0.58	0.97	1.11	0.61	1.08	1.34	0.78	1.08	1.02
$(r)$		0.95	0.93	0.0	0.71	0.85	0.88	0.72	0.88	0.91	0.80	0.88	0.87
Overall		1.05	2.30	14.51	15.61	41.98	12.54	32.10	41840.27	50289.35	0.82	281.01	4709.72

IHC - insertion height of the first cob; HEI – plant height; PCD - plant cob diameter; SCL - stripped cob length; SCD - stripped cob diameter; NRC - number of rows of grain per cob; NGR - number of grains per row; TNC – total number of cobs; FPS - final plant stand; NCP - number of cobs per plant; MTS - mass of one thousand seeds; YIE – yield. Genotypic variance ( $\sigma_g^2$ ), residual variance ( $\sigma_e^2$ ), phenotypic variance ( $\sigma_f^2$ ), heritability ( $\hat{h}_g^2$ ), coefficient of genotypic variation ( $CV_{gi}\%$ ), coefficient of residual variation ( $CV_e\%$ ), coefficient of relative variation, considering ( $CV_{gi}/CV_e$ ), coefficient of correlation ( $r$ ) and mean. \*, \*\* and ns = significant at 5 and 1% probability and no significant, respectively.



The phenotypic and genotypic correlations between the number of rows of grains per cob and mass of one thousand seeds were predominantly significant and negative, however, of less than (-0.60) magnitude (Table 3). The larger number of rows in a cob tends to reduce the grain size, thereby influencing the seed weight in a negative way. Cob volume and grain volume were those that most contributors to increasing the mean cob weight in production components, for super sweet maize populations [10].

The simultaneous selection of traits, such as number of grains per cob and grain weight per cob, is a difficult task for plant breeders, considering that the genes on which these traits are conditioned, have often negative correlations in genotypes [33]. The significance between the stripped cob diameter and the mass of one thousand seeds in the phenotypic and genotypic correlations, with values of 0.70 and 0.77, respectively, demonstrate that large cobs have a larger number of rows; nevertheless, the cob size is a relevant trait for a larger number of grains, contributing to a greater grain weight per cobs.

The variable insertion height of the first cob showed a positive phenotypic and genotypic correlation coefficient with regard to yield, with 0.31 and 0.29 respectively (Table 3). As stated by [34], when the correlation coefficient is positive, but the direct effect is negative, or when the value is insignificant, indirect effects cause the correlation. The variable stripped cob diameter also showed positive values by means of phenotypic and genotypic correlations, of 0.42 and 0.36, respectively, in relation to yield.

Table 3. Phenotypic ( $r_p$ ), genotypic ( $r_g$ ), and environmental ( $r_e$ ) correlations among 12 phytotechnical traits of 16 maize accessions in the municipality of Alegre, Espírito Santo State, Brazil, 2017

Variables	Correlation	HEI	SCL	SCD	NRC	NGR	NTE	TNC	NCP	MTS	YIE
IHC	( $r_p$ )	0.85**	-0.05	-0.25	-0.45	0.25	0.55+	0.51*	0.41	0.09	0.31
	( $r_g$ )	0.88++	-0.19	-0.32	-0.54	0.28	0.58+	0.53+	0.47	0.09	0.29
	( $r_e$ )	0.62++	0.33+	0.06	0.02	0.29+	0.39+	0.38+	0.28+	0.08	0.47++
HEI	( $r_p$ )		0.33	-0.00	-0.35	0.40	0.36	0.37	0.24	0.27	0.35
	( $r_g$ )		0.40	-0.03	-0.45	0.50	0.41	0.40	0.3	0.33	0.37
	( $r_e$ )		0.26	0.15	0.12	0.26	0.15	0.16	0.10	0.00	0.28+
SCL	( $r_p$ )			0.50*	0.11	0.33	-0.09	-0.15	0.01	0.50*	0.32
	( $r_g$ )			0.59	0.22	-0.07	-0.20	-0.26	-0.05	0.59	0.20
	( $r_e$ )			0.4++	-0.08	0.77++	0.09	0.06	0.11	0.39++	0.57++
SCD	( $r_p$ )				0.02	0.19	-0.05	0.13	-0.18	0.70**	0.42
	( $r_g$ )				-0.08	0.24	-0.14	0.08	-0.30	0.77++	0.36

	( $r_e$ )	0.40+	0.12	0.21	0.31	0.08	0.47++	0.61++
	( $r_p$ )		-0.26	-0.28	-0.13	-0.33	-0.55*	-0.19
NRC	( $r_g$ )		-0.49	-0.38	-0.18	-0.46	-0.57*	-0.32
	( $r_e$ )		0.15	0.05	0.09	0.00	-0.46++	0.25
	( $r_p$ )			0.48	0.30	0.51*	0.19	0.67**
NGR	( $r_g$ )			0.79+	0.5	0.88++	0.35	0.85+
	( $r_e$ )			-0.09	-0.10	-0.02	-0.09	0.38+
	( $r_p$ )				0.84**	0.85**	0.04	0.81**
TNC	( $r_g$ )				0.88++	0.86++	0.00	0.83++
	( $r_e$ )				0.68++	0.88++	0.18	0.74++
	( $r_p$ )					0.44	0.09	0.74
FPS	( $r_g$ )					0.51	0.06	0.78++
	( $r_e$ )					0.26	0.23	0.58++
	( $r_p$ )						0.00	0.66**
NCP	( $r_g$ )						-0.02	0.69+
	( $r_e$ )						0.09	0.61++
	( $r_p$ )							0.43
MTS	( $r_g$ )							0.43
	( $r_e$ )							0.42++

HEI – plant height; SCL - stripped cob length; SCD - stripped cob diameter; NRC - number of rows of grain per cob; NGR - number of grains per row; TNC – total number of cobs; FPS - final plant stand; NCP - number of cobs per plant; MTS - mass of one thousand seeds; YIE – yield. \*, \*\* and ns = significant at 5 and 1% probability and no significant, respectively, by the t test. +, ++ = significant at 1 and 5%, respectively, by the bootstrap method with 5000 simulations.

Before performing the path analysis, a multi co-linearity analysis was conducted among the variables [21]. Genotypic correlation matrices were submitted to the diagnosis of multi co-linearity on the basis of the number of conditions. The elimination of the variables stem diameter of the plant and number of cobs per plant was required in view of the severe multi co-linearity, 107,826.62. For the remaining variables, insertion height of the first cob; plant height; stripped cob length; stripped cob diameter; number of rows of grains per cob; number of grains per row; total number of cobs; final plant stand; mass of one thousand seeds; and yield, the number of conditions was 742.19, which enabled classifying them from moderate to strong, not affecting the path analysis statistics. It was decided to use, in this study, only the variables with no collinearity, given that, in strong or severe multi co-

281 **linearity**, the variances associated with path coefficients are likely to reach  
282 high values [12].

283 Table 4 depicts the direct and indirect effects of the explanatory variables  
284 using grain yield as the main variable. The coefficient of determination ( $R^2$ ) in  
285 the path analysis model displayed a value of 0.9236 and residual effects  
286 lower than 0.0143. As such, the model showed the cause and effect  
287 relationship between the explanatory variables and grain yield. The  
288 satisfactory use of path coefficients is directly linked to the composition of  
289 causal diagrams, which should be listed to the most important variables in  
290 the expression of the main variable [35] The diagram applied enabled to  
291 explain 92.36% ( $R^2$ ) of the variation in grain yield (Table 4).

292 In accordance with [30], in five hybrid maize lines, the weight of 100 grains  
293 was the variable yield, which generated the greatest direct effect on grain  
294 yield, being the most indicated for indirect selection regarding yield. Greater  
295 direct effect and greater total correlation on grain yield point a great  
296 contribution to increase yield [36]. The highest values of direct effects on  
297 grain yield were seen for the mass of one thousand seeds, with 0.8079,  
298 followed by the final plant stand, with 0.5850, and total number of cobs, with  
299 0.4359, while the stripped cob diameter displayed negative direct effect and  
300 high magnitude, with (-0.5173), and positive phenotypic correlation of mean  
301 magnitude, 0.42, with the indirect effect being considered on the mass of one  
302 thousand seeds, 0.6266, in the conditions in which the experiment was  
303 carried out (Table 4).

304 The direct effect between the variable insertion height of the first cob and  
305 yield was negative, (-0.6416), while the phenotypic correlation was positive  
306 and with a mean magnitude of 0.31, which should take into account the  
307 indirect effect in the final plant stand 0.3150 to benefit from the gain in these  
308 two traits (Table 4). As stated by [23], the final plant stand is critical to  
309 increase the production of dry mass and commercial cobs, both directly and  
310 indirectly. This is not the case in [36], who reported a positive value for the  
311 trait insertion height of the cobs in the harvest with the yield, even though  
312 they had worked with precocious cycle maize cultivars. In this case, a direct  
313 selection on the causal factor effect may not be efficient to improve the trait  
314 yield. Also in agreement with [37], the plant height in the harvest can be  
315 considered for indirect selection, because of the positive linear relationship  
316 with yield.

317 The low correlation observed between the insertion height of the first cob and  
318 the stripped cob diameter with the grain yield was caused by the negative  
319 indirect effects via the stripped cob length, number of rows of grains per cob,  
320 plant height, and total number of cobs, with very low values similar to the  
321 ones reported by [23]. Conversely, when a direct effect displays positive

322 results, indirect effects are responsible for the lack of correlation [38]. In this  
 323 way, for the indirect selection of more productive genotypes, the trait yield  
 324 should be related, with positive effects, through the desired trait. As  
 325 mentioned by [11], the negative correlation between different phenotypic  
 326 traits is assigned to different genes that are controlling these traits such is the  
 327 existing negative correlation between grains per cob and grain thickness.

328 For indirect effects with negative values, the simultaneous selection in an  
 329 indirect way leads to a low efficiency, [36]. In contrast, the positive values for  
 330 the direct effects of primary components on grain yield are good predictors of  
 331 genetic correlation according [39] and [40].

332 The indirect selection for increasing of the traits number of grains and  
 333 number of rows is efficient in increasing the grain weight [33]. Indirect  
 334 selection may result in faster genetic progress than direct selection of the  
 335 desired trait, [41] In an experiment with open-pollinated varieties, [31]  
 336 achieved a greater correlation between the number of grains per row and  
 337 yield 0.586.

338 Table 4. Direct and indirect effects of agronomic variables of maize from 16  
 339 varieties of maize in the municipality of Alegre, Espírito Santo State, Brazil,  
 340 2017

Variable	Effect	Via	Coefficients
IHC	Direct	YIE	-0.6416
	Indirect	HEI	0.1744
		SCL	-0.0072
		SCD	0.1686
		NRC	-0.1190
		NGR	0.0688
		TNC	0.2558
		FPS	0.3150
		MTS	0.0796
	Total		0.2946
HEI	Direct	YIE	0.1979
	Indirect	AIHC	-0.5656
		SCL	0.0153
		SCD	0.0202
		NRC	-0.0999
		NGR	0.1227
		TNC	0.1791
		FPS	0.2390
		MTS	0.2696
	Total		0.3786

SCL	Direct	YIE	0.0382
		IHC	0.1224
	Indirect	HEI	0.0796
		SCD	-0.3068
		NRC	0.0489
		NGR	-0.0174
		TNC	-0.0885
		FPS	-0.1573
		MTS	0.4822
	Total		0.2015
SCD	Direct	YIE	-0.5173
		IHC	0.2091
	Indirect	HEI	-0.0077
		SCL	0.0226
		NRC	-0.0187
		NGR	0.0588
		TNC	-0.0641
		FPS	0.0508
		MTS	0.6266
	Total		0.3602
NCR	Direct	YIE	0.2202
		IHC	0.3468
	Indirect	HEI	-0.0898
		SCL	0.0084
		SCD	0.0440
		NGR	-0.1189
		NTE	-0.1659
		FPS	-0.1090
		MTS	-0.4641
	Total		-0.3283
NGR	Direct	YIE	0.2417
		IHC	-0.1827
	Indirect	HEI	0.1004
		SCL	-0.0027
		SCD	-0.1260
		NRC	-0.1083
		TNC	0.3478
		FPS	0.2943
		MTS	0.2894
	Total		0.8539
TNC	Direct	YIE	0.4359
	Indirect	IHC	-0.3765

		HEI	0.0813
		SCL	-0.0077
		SCD	0.0761
		NRC	-0.0838
		NGR	0.1928
		FPS	0.5153
		MTS	0.0046
	Total		0.838
	Direct	YIE	0.5850
		IHC	-0.3455
		HEI	0.0808
		SCL	-0.0102
FPS	Indirect	SCD	-0.0449
		NRC	-0.0410
		NGR	0.1216
		TNC	0.3840
		MTS	0.0543
	Total		0.7842
	Direct	YIE	0.8079
		IHC	-0.0632
		HEI	0.0660
		SCL	0.0228
MTS	Indirect	SCD	-0.4012
		NRC	-0.1265
		NGR	0.0865
		TNC	0.0024
		FPS	0.0393
	Total		0.4342
	Coefficient of Determination		0.9632
	Effect of residual variable		0.0143

IHC - insertion height of the first cob; HEI – plant height; SCL - stripped cob length; SCD - stripped cob diameter; ; NRC - number of rows of grain per cob; NGR - number of grains per row; TNC – total number of cobs; FPS – final plant stand; MTS - mass of one thousand seeds; YIE – yield.

#### 4. CONCLUSION

The genetic variability identified in local breeds **allowed** the selection of genotypes to be used in genetic breeding programs, given that the traits under evaluation showed a genotypic correlation of greater magnitude than the phenotypic one, indicating genetic effects whether of an additive,



352 epistatic or dominance nature, exceeding the contribution of environmental  
353 origin.

354 The gain in the traits stripped cob diameter, which presented an indirect  
355 effect with mass of one thousand seeds, positive and with high magnitude,  
356 as well as the indirect effect of the variable insertion height of the first cob in  
357 the final plant stand, should be exploited. The final plant stand and the mass  
358 of one thousand seeds **were** determinants to directly increase the grain yield.

359

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368

## 369 **COMPETING INTERESTS**

370

371 Authors have declared that no competing interests exist in this study.

372

## 373 **AUTHORS' CONTRIBUTIONS**

374

375 “Author Richardson Sales Rocha” designed the study and performed the  
376 statistical analysis, “Author Maxwell Rodrigues Nascimento” managed the  
377 analyses of the study, “Author José Tiago Barroso Chagas” and “Author  
378 Rafael Nunes De Almeida” managed the literature searches, “Author Paulo  
379 Ricardo Dos Santos” and “Author Camila Queiroz Da Silva Sanfim De  
380 Sant’anna” supporting at the experimental design and traits evaluations,  
381 “Author Derivaldo Pureza Da Cruz” and “Author Kleyton Danilo Da Silva  
382 Costa” suggests path analysis, “Author Geraldo De Amaral Gravina” and  
383 “Author Rogério Figueiredo Daher” review the manuscript and suggests  
384 some alterations. All authors read and approved the final manuscript”.

385

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