

2 **Association Among Agro-morphological Traits by**
3 **Correlations and Path in Selection of Maize**
4 **Genotypes**

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9 **ABSTRACT**

10 In the State of Espírito Santo (Brazil), 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. The study aimed to analyze the cause and effect of associations between agro-morphological traits for increasing yields in maize populations cultivated. A total of 16 maize accessions of in vivo conservation works in eight different locations 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, accessions presented enough genetic variability, showing significant differences for all traits analyzed statistically ($P \leq 0.05$). The first cob insertion height with plant height and final plant stand variables presented heritability (h^2) higher than 80%, suggesting that superior genotypes could be determined. Highest estimates of phenotypic correlations (r_p) were found between the first cob insertion height with plant height and total number of cobs with number of cobs per plant (0.85). For estimates of genotypic correlations (r_g), the highest were between (i) the first cob insertion height and plant height; (ii) number of grains per row and number of cobs per plant; (iii) total number of cobs and final plant stand. Maize final plant stand and one thousand seed weight were observed to be traits that could be determinant in grain yield increase.

11
12 **Keywords:** *Zea mays* L., heritability, genetic parameter, open-pollination, in vivo
13 germplasm, simultaneous selection.

14
15 **1. INTRODUCTION**

16
17 Maize (*Zea mays* L.) is one of the most grown cereals worldwide, for its high
18 genetic diversity and wide adaptability [1]. Such a fact, associated with its
19 nutritional characteristics, makes maize also one of the cereals of
20 commercial importance in different countries [2]. Brazil is the third largest
21 maize producer, meaning that the estimated national production for the first
22 2018/19 harvest is 27.4 million tons and 63.7 million tons for the second one,
23 with a mean yield of 5.2 t ha^{-1} in the last harvest [3].

24 In the State of Espírito Santo, family farmers have grown a number of maize
25 varieties for decades, consisting of open-pollinated populations with valuable
26 importance for the livelihood of smallholder farmers. Estimated total
27 contribution for the 2019 harvest is 38.0 thousand t, with yield (approximately
28 2.8 t ha⁻¹) being one of the lowest in the country [3].

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

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

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

54 In the State of Espírito Santo (Brazil), both existing agronomic institutes (Ifes
55 and Incaper) have been working on studies with maize populations grown in
56 the State, gathering relevant germplasm and working on breeding strategies
57 so as to increase grain yield [15]; [16]. As such, studying correlations
58 between agronomic traits in these maize populations will contribute to the
59 optimization of the strategies used by breeders.

60 The purpose of this study was to analyze the cause and effect associations
61 of agronomic traits for increasing yields in maize populations grown in
62 Espírito Santo.

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64 **2. MATERIAL AND METHODS**

65

66 The experiment was conducted in the Federal Institute of Education, Science
 67 and Technology of Alegre in 2017. According to the Köppen international
 68 classification, the climate of the region is tropical wet and hot with, one the
 69 one hand, a cold and dry winter, and on the other hand, a hot and rainy
 70 summer. The mean temperature in the region is 23.1°C, and the mean
 71 annual precipitation is 1,341 mm [17].

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

77 Table 1 - Maize accessions (genotypes) evaluated in accordance with
 78 agronomic traits, in Alegre (Brazil), and their locations of origin.

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

79 GEBs= Germplasm banks; IFES= Federal Institute of Espírito Santo;
 80 INCAPER= Institute of Technical Assistance and Rural Development.

81 In order to implement the evaluation test of the genotypes in the field, a
 82 randomized complete block design with sixteen genotypes (treatments) and
 83 three replicates was used according to the following statistical model:
 84 $X_{ij} = m + t_i + b_j + e_{ij}$ in which: m = represents the overall mean, t_i =
 85 treatment effect (genotypes), b_j = block effect, e_{ij} = effect of the experimental
 86 error. The experiment comprised 48 experimental units, each of them being

87 composed of three lines of 4.0 m in length with 1.0 m of inter-row spacing.
88 The evaluation of agronomic traits was performed on 16 plants within the
89 central line of the plot.

90 During sowing 15 seeds per linear meter were uniformly distributed in furrow.
91 Twenty one days after sowing (DAS), thinning was performed to establish a
92 population of 5 plants per linear meter of furrow, corresponding to a density
93 of 50,000 plants ha⁻¹ (adapted to Corrêa *et al.*, 2014). Standard maize
94 cultivation practices were followed by pre-emergence herbicide application
95 [18]. The maize grain was naturally dried on plant, without using any
96 desiccant, until it reached 13% moisture content. Manual harvest occurred in
97 February 2018.

98 Twelve different agronomic traits such as Plant height (HEI), first cob
99 Insertion height (CIH) and Total number of cobs (TNC) were evaluated. HEI
100 was measured from the base to the last flag leaf before tasselling and CIH
101 from the base of the stem to the first cob. All traits under evaluation are
102 descriptors established by Biodiversity International [19].

103 Aside from conventional treatments to control the fall armyworm (*Spodoptera*
104 *frugiperda*), leaf sprays were carried out using the Dipel WP biological
105 insecticide, made from *Bacillus thuringiensis*. The quantity used was of 500g
106 ha⁻¹ and the applications were made with the aid of a manual knapsack
107 sprayer with a 20 litre capacity. Mechanical weeding was used to control the
108 weeds at 30 DAS.

109 For statistical analyses, the phenotypic (r_p), genotypic (r_g), and environmental
110 (r_e) correlation analyses and linear regression analysis were performed. For
111 correlations, the following expressions were applied: phenotypic correlations:

112 $r_F = \frac{MPG_{XY}}{\sqrt{MSG_X MSG_Y}}$; genotypic correlations: $r_G = \frac{(MPG_{XY} - PMR_{XY})/r}{\sqrt{\hat{\Phi}_{g(X)} \hat{\Phi}_{g(Y)}}} = \frac{\hat{\Phi}_{g(XY)}}{\sqrt{\hat{\Phi}_{g(X)} \hat{\Phi}_{g(Y)}}}$,

113 and environmental correlations: $r_p = \frac{MPG_{XY}}{\sqrt{MSR_X MSR_Y}}$. in which, MPG_x = mean

114 product between genotypes for traits X and Y; MPR_{xy} = mean product
115 between residues for traits; MSG_x = mean square between genotypes for
116 trait X; MSG_y = mean square between genotypes for trait Y; MSR_x = mean
117 square between residues for trait X; MSR_y = mean square between residues
118 for trait Y; $\hat{\Phi}_{g(XY)}$ = genotypic covariance estimator; $\hat{\Phi}_{g(X)}$, $\hat{\Phi}_{g(Y)}$ = estimators
119 of quadratic components associated with genotypic variabilities for traits X
120 and Y, respectively.

121

122 The path analysis consisted of studying the direct and indirect effects of the
123 above-mentioned explanatory independent variables (X) on grain yield, main
124 dependent variable (Y). Considering Y to be a complex trait, resulting from
125 the combined action of other traits, the following model can be defined:

126 $Y = \beta_{1X_1} + \beta_{2X_2} + \dots + \beta_{nX_n} + \epsilon$, in which: X_1, X_2, \dots, X_n are the explanatory
127 variables, and Y is the main variable (or dependent variable). The direct and
128 indirect effects of the explanatory variables are estimated on the main
129 variable. Therefore, $r_{iy} = p_i + \sum_{j \neq 1}^n p_{ij} r_{ij}$ in which: correlation between the
130 main variable (Y) and the i -th explanatory variable; p_i : direct effect of variable
131 i on the main variable; and $p_j r_{ij}$: indirect effect of variable i by means of
132 variable j on the main variable.

133

134 The significance of genotypic correlation coefficient and b_1 of the regression
135 were evaluated by the "t" test, and the bootstrap with 5000 simulations for
136 phenotypic and environmental correlations was applied in accordance with
137 Ferreira *et al* [20].

138

139 In order to check the co-linearity between traits, a multi co-linearity test was
140 conducted, in line with Montgomery and Peck cited by Cruz *et al* [21].
141 Subsequently, it was carried out through a split of simple correlation
142 coefficients into direct and indirect effects provided by the trail analysis. For
143 all analyses, computational resources from Gene stat program were used
144 [22].

145

146 3. RESULTS AND DISCUSSION

147

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

153 As a general rule, the genotypes presented enough genetic variability,
154 showing a significant difference for all the traits under analysis, excluding the
155 stem diameter of the plant, by the F test ($P=.05$) (Table 2). In this regard,
156 Almeida *et al* [24] observed the variability of field maize, super sweet maize
157 and teosinte maize populations, pointing out that the low plant genetic
158 variability leads to lower genetic gain for breeding programs. Experimental
159 coefficients of variation ($CV_e\%$) showed magnitudes ranging from 5%, for
160 stripped cob diameter, to 25.14%, for plant stem diameter. In turn, the
161 genotypic coefficient variation ($CV_{gi}\%$) varied from 0.0 to 22.5% for traits like
162 plant stem diameter and yield, respectively. These values are basically
163 explained by the variability of the genetic material used in the analysis.
164 Research on other crops has shown that high variability among genotypes
165 facilitated the selection process of superior material [25-28].

166

167 As reported by Falconer [29], heritability values (h^2) above 80% and variation
168 index above unit 1.0 ensure satisfactory selection gains. For the insertion

169 height of the first cob, plant height, and final plant stand, this situation was
170 verified suggesting that, for these variables, superior genotypes can be
171 obtained by means of simple selection methods, such as mass selection
172 (Table 3).

173 Among 55 pairs of combinations for 12 variables evaluated, 13 had
174 significant (r_p), 12 to 1 or 5% probability by the "t" test, and only one
175 significant to 1% by the bootstrap method with 5000 simulations (Table 2).
176 There was a positive and high variation from 0.50 to 0.85 for 12 pairs. (r_g)
177 also presented 13 significant pairs, but 12 by the bootstrap method with 5000
178 simulations at 1 or 5% probability, and only one to 5% probability by the "t"
179 test. For this one, there was a positive and high variation from 0.53 to 0.88
180 for 12 pairs. For environmental correlations (r_e), 23 combinations were
181 significant using the bootstrap method with 5000 simulations at 1 or 5%
182 probability, varying from 0.4 to 0.88 positive pairs (Table 3).

183 The highest estimate of ($r_p = 0.85$) was noticed between the first cob
184 insertion height and the total number of cobs with number of cobs per plant.
185 Lowest correlations (r_p) were observed between stripped cob length with
186 stripped cob diameter 0.50 and stripped cob length and mass of one
187 thousand seeds 0.50, being significant estimates by the "t" test. These
188 results corroborate the ones reported by Souza, *et al* [30], which indicate
189 variable cob height as the most influential in yield variation and of greater
190 relevance in indirect selection for yield. Number of rows of grains per cob and
191 mass of one thousand seeds showed significant and high value, though
192 negative ($r_p = -0.55$). The higher the number of rows, the lower the
193 conversion of photoassimilates into starch [31].

194 Among estimates of (r_g), the highest one was 0.88 between the first cob
195 insertion height and plant height. Within positive estimates, the lowest one
196 was found between the first cob insertion height and final plant stand, with
197 0.53. There was only one significant and high pair, although with a negative
198 value ($r_g = -0.57$) between number of rows of grains per cob and mass of one
199 thousand grains. The genotype selection with a larger diameter and one
200 thousand grain weight enables to obtain genotypes with a good yield [32].
201 Moreover, regarding (r_e), the highest estimate happened between total
202 number of cobs and number of cobs per plant, with 0.88. Stripped cob length
203 and stripped cob diameter 0.4 was the lowest positive and significant value.
204 There was a significant and high pair, however, with a negative value
205 between the number of grain rows per cob and one thousand seed weight
206 ($r_g = -0.46$), (Table 3).

207 The first cob insertion height indicated negative correlations (r_g) with stripped
208 cob length of (-0.19), stripped cob diameter of (-0.32), and number of rows of
209 grains per cob of (-0.54). Positive correlations were observed with number of

210 grains per row 0.28, number of cobs per plant 0.47, one thousand seed
211 weight 0.09 and yield 0.29, but still not significant in all cases (Table 3). In
212 accordance with Cruz *et al* [21], a non-significant or low magnitude
213 correlation coefficient does not suggest lack of relationship between two
214 variables, but absence of a linear relationship between them.

UNDER PEER REVIEW

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Table 2. Estimates of genetic parameters for agro-morphological traits of 16 maize genotypes in Alegre, Brazil.

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												
(σ_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
(σ_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
(σ_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

218 IHC - insertion height of the first cob; HEI – plant height; PCD - plant cob diameter; SCL - stripped cob length; SCD -
219 stripped cob diameter; NRC - number of rows of grain per cob; NGR - number of grains per row; TNC – total number
220 of cobs; FPS - final plant stand; NCP - number of cobs per plant; MTS - mass of one thousand seeds; YIE – yield.
221 Genotypic variance (σ_g^2), residual variance (σ_e^2), phenotypic variance (σ_f^2), heritability (\hat{h}_g^2), coefficient of genotypic
222 variation ($CV_{gi}\%$), coefficient of residual variation ($CV_e\%$), coefficient of relative variation, considering (CV_{gi}/CV_e),
223 coefficient of correlation (r) and mean. *, ** and ns = significant at 5 and 1% probability and non-significant,
224 respectively.
225

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

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

241 The first cob insertion height showed positive phenotypic and genotypic
 242 correlation coefficients regarding maize yield, with 0.31 and 0.29 respectively
 243 (Table 3). As stated Lorentz *et al* [34], when the correlation coefficient is
 244 positive, but the direct effect is negative, or when the value is insignificant,
 245 indirect effects cause the correlation. The variable stripped cob diameter also
 246 showed positive values by means of phenotypic and genotypic correlations,
 247 of 0.42 and 0.36, respectively, in relation to yield.

248 Table 3. Phenotypic (r_p), genotypic (r_g), and environmental (r_e) correlations
 249 among 12 agro-morphological traits of 16 maize accessions in Alegre, Brazil.

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++
NRC	(r_p)					-0.26	-0.28	-0.13	-0.33	-0.55*	-0.19

	(r _g)					-0.57*	-0.32
	(r _e)	-0.49	-0.38	-0.18	-0.46	-0.46++	0.25
	(r _p)					0.19	0.67**
NGR	(r _g)		0.48	0.30	0.51*	0.35	0.85+
	(r _e)		0.79+	0.5	0.88++	-0.09	0.38+
	(r _p)		-0.09	-0.10	-0.02	0.04	0.81**
TNC	(r _g)			0.84**	0.85**	0.00	0.83++
	(r _e)			0.88++	0.86++	0.18	0.74++
	(r _p)			0.68++	0.88++	0.09	0.74
FPS	(r _g)				0.44	0.06	0.78++
	(r _e)				0.51	0.23	0.58++
	(r _p)				0.26	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++

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

257 Before performing the path analysis, a multi co-linearity analysis was
 258 conducted among the variables [21]. Genotypic correlation matrices were
 259 submitted to the diagnosis of multi co-linearity on the basis of the number of
 260 conditions. The elimination of the variables stem diameter of the plant and
 261 number of cobs per plant was required in view of the severe multi co-
 262 linearity, 107,826.62. For the remaining variables, insertion height of the first
 263 cob; plant height; stripped cob length; stripped cob diameter; number of rows
 264 of grains per cob; number of grains per row; total number of cobs; final plant
 265 stand; one thousand seed weight; and yield, the number of conditions was
 266 742.19, which enabled classifying them from moderate to strong, not
 267 affecting the path analysis statistics. It was decided to use, in this study, only
 268 the variables with no collinearity, given that, in strong or severe multi co-
 269 linearity, the variances associated with path coefficients are likely to reach
 270 high values [12].

271 Table 4 depicts the direct and indirect effects of the explanatory variables
272 using grain yield as the main variable. The coefficient of determination (R^2) in
273 the path analysis model displayed a value of 0.92 and residual effects lower
274 than 0.014. As such, the model showed the cause and effect relationship
275 between the explanatory variables and grain yield. The satisfactory use of
276 path coefficients is directly linked to the composition of causal diagrams,
277 which should be listed to the most important variables in the expression of
278 the main variable [35]. The diagram applied enabled to explain 92% (R^2) of
279 the variation in grain yield (Table 4).

280 In accordance with Souza *et al* [30], in five hybrid maize lines, the weight of
281 100 grains was the variable yield, which generated the greatest direct effect
282 on grain yield, being the most indicated for indirect selection regarding yield.
283 Greater direct effect and greater total correlation on grain yield point a great
284 contribution to increase yield [36]. The highest values of direct effects on
285 grain yield were seen for one thousand seed weight, with 0.81, followed by
286 the final plant stand, with 0.58, and total number of cobs, with 0.43, while the
287 stripped cob diameter displayed negative direct effect and high magnitude,
288 with (-0.52), and positive phenotypic correlation of mean magnitude, 0.42,
289 with the indirect effect being considered on the mass of one thousand seeds,
290 0.63, in the conditions in which the experiment was carried out (Table 4).

291 The direct effect between the variable insertion height of the first cob and
292 yield was negative, (-0.64), while the phenotypic correlation was positive and
293 with a mean magnitude of 0.31, which should take into account the indirect
294 effect in the final plant stand 0.31 to benefit from the gain in these two traits
295 (Table 4). As stated by Nascimento *et al* [23], the final plant stand is critical to
296 increase the production of dry mass and commercial cobs, both directly and
297 indirectly. This is not the case in Kleinpaul *et al* [37], who reported a positive
298 value for the trait insertion height of the cobs in the harvest with the yield,
299 even though they had worked with precocious cycle maize cultivars. In this
300 case, a direct selection on the causal factor effect may not be efficient to
301 improve the trait yield. Also in agreement with Kleinpaul *et al* [37], the plant
302 height in the harvest can be considered for indirect selection, because of the
303 positive linear relationship with yield.

304 The low correlation observed between the insertion height of the first cob and
305 the stripped cob diameter with the grain yield was caused by the negative
306 indirect effects via the stripped cob length, number of rows of grains per cob,
307 plant height, and total number of cobs, with very low values similar to the
308 ones reported by [23]. Conversely, when a direct effect displays positive
309 results, indirect effects are responsible for the lack of correlation [38]. In this
310 way, for the indirect selection of more productive genotypes, the trait yield
311 should be related, with positive effects, through the desired trait. As
312 mentioned by Nemati *et al* [11], the negative correlation between different

313 phenotypic traits is assigned to different genes that are controlling these
 314 traits such is the existing negative correlation between grains per cob and
 315 grain thickness.

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

320 The indirect selection for increasing of the traits number of grains and
 321 number of rows is efficient in increasing the grain weight [33]. Indirect
 322 selection may result in faster genetic progress than direct selection of the
 323 desired trait, [41] In an experiment with open-pollinated varieties, Balbinot Jr
 324 *et al* [31] achieved a greater correlation between the number of grains per
 325 row and yield 0.586.

326 Table 4. Direct and indirect effects of agro-morphological traits of 16 maize
 327 varieties in Alegre, Brazil.

Variable	Effect	Via	Coefficients
IHC	Direct	YIE	-0.6416
		HEI	0.1744
	Indirect	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
		AIHC	-0.5656
	Indirect	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	
		Direct	YIE	-0.5173
			IHC	0.2091
			HEI	-0.0077
			SCL	0.0226
	SCD	Indirect	NRC	-0.0187
			NGR	0.0588
			TNC	-0.0641
			FPS	0.0508
			MTS	0.6266
		Total	0.3602	
		Direct	YIE	0.2202
			IHC	0.3468
			HEI	-0.0898
			SCL	0.0084
	NCR	Indirect	SCD	0.0440
			NGR	-0.1189
			NTE	-0.1659
			FPS	-0.1090
			MTS	-0.4641
		Total	-0.3283	
		Direct	YIE	0.2417
			IHC	-0.1827
			HEI	0.1004
			SCL	-0.0027
	NGR	Indirect	SCD	-0.1260
			NRC	-0.1083
			TNC	0.3478
			FPS	0.2943
			MTS	0.2894
		Total	0.8539	
		Direct	YIE	0.4359
			IHC	-0.3765
			HEI	0.0813
	TNC	Indirect	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	

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

332

333 4. CONCLUSION

334

335 The genetic variability identified in local cultivars allowed the selection of
336 maize genotypes to be used in breeding programs. Agro-morphological traits
337 evaluated showed genotypic correlations of greater magnitude than the
338 phenotypic ones, indicating genetic different effects of additive, epistatic or
339 dominance nature. These effects exceeded variations of environmental
340 origin.

341 A gain of high magnitude on traits like stripped cob diameter which presented
342 an indirect effect with one thousand seed weight as well as the indirect effect
343 of first cob insertion height in final plant height, should be exploited. The final

344 plant stand in maize and one thousand seed weight were determinant in
345 grain yield increase.

346

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355

356 **COMPETING INTERESTS**

357

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

359

360 **AUTHORS' CONTRIBUTIONS**

361

362 “Author Richardson Sales Rocha” designed the study and performed the
363 statistical analysis, “Author Maxwell Rodrigues Nascimento” managed the
364 analyses of the study, “Author José Tiago Barroso Chagas” and “Author
365 Rafael Nunes De Almeida” managed the literature searches, “Author Paulo
366 Ricardo Dos Santos” and “Author Camila Queiroz Da Silva Sanfim De
367 Sant’anna” supporting at the experimental design and traits evaluations,
368 “Author Derivaldo Pureza Da Cruz” and “Author Kleyton Danilo Da Silva
369 Costa” suggests path analysis, “Author Geraldo De Amaral Gravina” and
370 “Author Rogério Figueiredo Daher” review the manuscript and suggests
371 some alterations. All authors read and approved the final manuscript”.

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