# ADAPTABILITY AND STABILITY FOR IRON AND ZINC IN COWPEA BY AMMI ANALYSIS<sup>1</sup>

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**ABSTRACT** - Iron and zinc deficiency is one of the main problems affecting vulnerable populations in the Colombian Caribbean, thereby generating malnutrition from the consumption of foods with low content of essential minerals. The objective of this study was to evaluate the genotype-environment interaction for iron and zinc accumulation in grains in 10 cowpea bean genotypes by additive main effects and multiplicative interaction (AMMI) model and to select the most stable ones to stimulate their planting or as parents in the genetic improvement program. Nine promising lines and a commercial control were evaluated using the randomized complete block design with 10 treatments and four replications in 10 environments of the northern Colombia in the second semester of 2017 and first of 2018. The adaptability and stability analysis was done using AMMI model. The results showed highly significant differences at the level of environments, genotypes, and genotype-environment interaction for iron and zinc, demostrating a differential adaptability of genotypes in the test environments. Genotypes 2 and 3 expressed greater adaptability and stability for iron contents in the seed; while genotype 1, recorded it for zinc contents. These three genotypes outperformed the commercial control and, therefore, can be recommended for planting or be used as parents in the genetic improvement program.

Keywords: Vigna unguiculata. Biofortification. Genotype-environment interaction. Micronutrients. Malnutrition.

# ADAPTABILIDADE E ESTABILIDADE DE FERRO E ZINCO EM FEIJÃO-CAUPI POR MEIO DA ANÁLISE AMMI

**RESUMO** - A deficiência de ferro e zinco é um dos principais problemas que afeta populações vulneráveis do Caribe colombiano, gerando desnutrição pelo consumo de alimentos com baixos conteúdos de minerais essenciais. O objetivo do estudo foi avaliar a interação genótipo x ambiente para acumulação de ferro e zinco nos grãos em 10 genótipos de feijão-caupi por meio do o modelo de efeitos principais aditivos e interação multiplicativa (AMMI) e selecionar os mais estáveis para estimular o plantio ou como parentais no programa de melhoramento genético. Nove linhas promissoras e um controle comercial foram avaliados usando o delineamento em blocos ao acaso com 10 tratamentos e quatro repetições em 10 ambientes do norte da Colômbia no segundo semestre de 2017 e primeiro de 2018. A análise de adaptabilidade e estabilidade foi realizada utilizando o modelo AMMI. Os resultados mostraram diferenças altamente significativas a nível de ambientes, genótipos nos ambientes de teste. Os genótipos 2 e 3 expressaram maior adaptabilidade e estabilidade e estabilidade para o teor de ferro nos grãos; enquanto o genótipo 1, registrou-o para o teor de zinco. Esses três genótipos superaram o controle comercial e, portanto, podem ser recomendados para o plantio ou serem usados como parentais no programa de melhoramento genético.

**Palavras-chave**: *Vigna unguiculata*. Biofortificação. Interação Genótipo-ambiente. Micronutrientes. Desnutrição.

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# **INTRODUCTION**

Cowpea (*Vigna unguiculata* (L.) Walp), is cultivated in tropical and subtropical areas of the world and for the year 2018, 12.496.305 ha were harvested with a production of 7.233.408 t (FAO, 2020). It is consumed as green and dry grain, and as fodder, for its protein, iron and zinc contents (MÁRQUEZ-QUIROZ et al., 2015; GERRANO et al., 2019), since they contribute significantly to mitigate the problems of hidden hunger, in vulnerable populations in both rural and urban areas of the Colombian Caribbean.

Mineral deficiencies in the populations of northern Colombia are related to frequent consumption of cereals and tubers with poor iron content, which affects the level of hemoglobin, and the transport of oxygen in the blood, DNA synthesis, metabolic processes and energy production (LIU et al., 2014; SINGH et al., 2016). On the other hand, zinc deficiency affects bone formation, the immune system, increases the susceptibility to developing cancer, and affects sexual functionality and maturation (LIU et al., 2014; GADDAMEEDI et al., 2018).

An alternative to mitigate the public health problems caused by hidden hunger (KUMAR et al., 2019), is the biofortification of crops (GUILLÉN-MOLINA et al., 2016), apart from other options such as crop fertilization, fortification of food products and mineral supplementation and vitamin (SINGHAL et al., 2018). Biofortification respects eating habits with economic and social advantages for people with lower incomes. For this reason, the genetic improvement of crops focuses on improving yield and nutritional quality to contribute to the improvement of the biochemical indicators of iron, zinc and other minerals.

Genotype-environment interaction affects quantitative characteristics such as the yield and the content of iron and zinc, as has been reported for mung beans (SINGH et al., 2013); cowpea (OLIVEIRA et al., 2017), lentil (DARAI et al., 2017); pearl millet (SINGHAL et al., 2018) and rice (INABANGAN-ASILO et al., 2019), becoming an obstacle to the plant breeder, in the identification and selection of stable genotypes for these two characteristics in the grain (DARAI et al., 2017).

Various methods have been used to measure the genotype-environment interaction for iron and zinc, including that of Eberhart and Russell (1966) by Singh et al. (2013), Kant's index by Inabangan-Asilo et al. (2019) and others. The AMMI method (Analysis of main additive effects and multiplicative interaction) has increasingly been applied in agricultural species (DARAI et al., 2017; OLIVEIRA et al., 2017; SINGHAL et al., 2018; INABANGAN-ASILO et al., 2019), because it allows characterizing and grouping both genotypes and environments and analyzes genotypeenvironment interactions with greater efficiency (ZOBEL; WRIGHT; GAUCH, 1988) and its interpretation is facilitated by biplot charts. Therefore, the objective of this research was to evaluate the behavior of lines previously selected for high content of iron and zinc, in contrasting environments of the Colombian Caribbean to know their interaction by AMMI analysis and select the most stable genotypes for commercial planting or use as a parent.

# MATERIALS AND METHODS

## Genetic material and experimental location

Nine advanced lines of the genetic plant breeding program of the Universidad de Córdoba, Colombia, previously selected for their iron and zinc content, were used: 1. LC-029-16; 2. LC-002-016; 3. LC-036-016; 4. LC-009-016; 5. LC-021-016; 6. L-019; 7. LC-006-016; 8. LC-005-016; 9. L-014-016. and as a control 10. CAUPICOR. These lines were evaluated in 10 environments in the Caribbean region of Colombia in semesters second (B) of 2017 and first (A) of 2018, identified as follows: 1. Cereté -Córdoba (CE7B), 2. Mahates-Bolívar (MA7B), 3. Montería-Córdoba (MO7B), 4. Polonuevo-Atlántico (PN7B), 5. Sampués-Sucre (SA7B), 6. Cereté-Córdoba (CE8A), 7. Leticia-Bolívar (LE8A), 8. Polonuevo-Atlántico (PN8A), 9. Sampués-Sucre (SA8A), and 10. Villanueva-Guajira (VI8A). Soils differ from one subregion to another, with fertility ranging from high to very low, depending on precipitation and the influence or not of rivers and tributaries. The experiments were carried out without edaphic fertilization or supplementary irrigation. The climate is tropical, a typical savanna subtype Aw, according to the Köppen classification, with a temperature variation between 24 and 28 °C and annual rainfall between 800 and 1800 mm.

## Experimental design and statistical analysis

In each environment the experimental design of complete random blocks with 10 treatments and four replications was used. The experimental unit consisted of plots of six rows of five meters long, spaced at 0.80 m and 0.40 m between plants for a population of 31.250 plants ha<sup>-1</sup>.

Determinations of the iron and zinc contents were made at the International Center for Tropical Agriculture (CIAT), Cali-Colombia. For this, 5 grams of seed were taken per treatment at harvest.

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The iron and zinc contents, in mg kg-<sup>1</sup>, was determined by atomic absorption spectroscopy using a Solaar Unicam 969 kit.

The analysis of variance for estimation of the main effects, and the effect of the genotypeenvironment interaction (GEI) explained by principal components was performed through the AMMI model (additive main effects and multiplicative interaction model). For this model, the effects of genotypes and environments are additive and linear, while GEI has multiplicative effects explained by principal component analysis (ZOBEL; WRIGHT; GAUCH, 1988; EBDON; GAUCH JR, 2002). The model is as follows:

$$Y_{ger} = \mu + \alpha_g + \beta_e + \sum_{n=1}^{V} \lambda_n \zeta_{gn} \eta_{en} + \varepsilon_{ger}$$

The meaning of each model term is:  $Y_{ijk}$  is the performance of genotype g in environment e for repeat r;  $\mu$  is the general mean;  $\alpha_g$  is the effect of genotype g;  $\beta_e$  is the effect of the environment e and; N is the number of axes in the principal component analysis (PCA) retained in the model;  $\lambda_n$  is the eigenvalue for the n axis of the PCA;  $\varsigma_{gn}$  and  $\eta_{en}$  are the PCA scores for their own axes; and  $\varepsilon_{ger}$  is the model error.

PCA scores for environments and genotypes in the interaction have units equivalent to the square root of the response variable (EBDON; GAUCH JR, 2002). For the AMMI analysis the GEA-R software (Genotype x Environment Analysis with R for Windows) Version 4.1 of the International Center for Maize and Wheat Improvement CIMMYT (PACHECO et al., 2015) was used.

## **RESULTS AND DISCUSSION**

Estimates of the mean squares of the main additive effects of environment (E), genotype (G) and genotype-environment interaction (GEI) for the iron and zinc contents (Table 1) are highly significant (p < 0.01). This indicates that these minerals vary in genotypes according to the environment where they were cultivated, there is also variation between them and this variation is not independent, but each genotype accumulates in the grain different amounts according to the environment where it is cultivated, as has been evidenced in previous studies carried out by Muranaka et al. (2016), Oliveira et al. (2017), and Singhal et al. (2018). It is necessary to examine the GEI to know its magnitude and determine whether it is positive or negative in order to be successful in the selection and recommendation of cultivars for specific environments or multi-environments, given the importance of these micronutrients in human health, and to make a better use of the environmental supply (BASHIR et al., 2014; SINGHAL et al., 2018; SANTOS et al., 2019).

The sums of squares indicate that environments (E) retain 75.84% and 90.75%, respectively, of the total variation of the AMMI model for both nutrients; secondly, the GEI with 18.50% and 6.97% and then, genotypes (G), with 5.66% and 2.28%, respectively, which is consistent with research on cowpea beans by Santos et al. (2015) and lentil by Darai et al. (2017).

The decomposition of the GEI into principal components and the Gollob test (1968) shows that, for iron content, the first two principal components (PC1 and PC2) were highly significant, the third (PC3) significant and together explain 77.22% of the sum of squares of the interaction with 45 of 81 degrees of freedom (55.6%); the rest of the principal components are not significant, which in a first analysis constitutes the residual of the interaction. However, when the variation (sums of squares) includes components from the third (PC3) to the ninth (PC9) it is estimated as 1397.88 with 49 degrees of freedom, equivalent to a mean square of 28.53 that is not significant, with F-test value of 1.37 as can be deduced from Table 1. This indicates that the first two principal components explain the GEI efficiently, since they represent 93.51% of the AMMI model, and the residual of the interaction is only 6.49%.

For zinc content, despite the fact that the first two principal components were highly significant, the sums of squares from the second principal component to the ninth (PC2 to PC9) and the associated 64 degrees of freedom, allow the estimation of a CM of 8.24, which is not significant with an F value of 1.24. This indicates that, for the zinc content, the interaction is effectively explained with only the first principal component (PC1) that represents 95.86% of the AMMI model and the GEI residual only 4.14%.

The biplot in which the abscissa shows the main effects (means of the genotypes and environments for the iron content) and ordered the first axis of the principal component analysis (PCA) of the AMMI model is presented in Figure 1. This graphical representation contains 81.50% of the sum of squares of the model and 42.74% of the sum of squares of the GEI and constitutes AMMI1, that is, the reduced model that includes only the first axis of the PCA.

SV	DF	Iron									
		SC	%SS	%SSA	MS	F	SS	%SS	%SSA	MS	F
Е	9	16344.2	75.8	75.8	1816.0**	87.5	11567.2	90.8	90.8	1285.2**	194.0
G	9	1220.1	5.7	81.5	135.6**	6.5	291.0	2.3	93.0	32.3**	4.9
GEI	81	3987.6	18.5	100.0	49.2**	2.4	888.0	7.0	100.0	11.0**	1.7
PC1	17	1704.1	42.7	42.7	100.2**	4.9	360.4	40.6	40.6	21.2**	3.3
PC2	15	885.6	22.2	64.9	59.0**	2.9	241.3	27.2	67.8	16.1**	2.5
PC3	13	489.3	12.3	77.2	37.6*	1.8	120.6	13.6	81.4	9.3 ns	1.4
PC4	11	368.5	9.2	86.5	33.5 ns	1.6	73.6	8.3	89.6	6.7 ns	1.0
PC5	9	252.7	6.3	92.8	28.1 ns	1.4	51.6	5.8	95.5	5.7 ns	0.9
PC6	7	173.5	4.4	97.2	24.8 ns	1.2	22.6	2.6	98.0	3.2 ns	0.5
PC7	5	109.7	2.8	99.9	21.9 ns	1.1	10.2	1.2	99.1	2.0 ns	0.3
PC8	3	4.0	0.1	100.0	1.3 ns	0.1	5.9	0.7	99.8	2.0 ns	0.3
PC9	1	0.2	0.0	100.0	0.2 ns	0.01	1.8	0.2	100.0	1.8 ns	0.3
PC10		0	0	100.0	0	0.00	0	0	100.0	0	0
Error	300	6227.88	0	0	20.8		1987.44	0	0	6.62	

**Table 1**. Analysis of variance and Gollob's test of the AMMI Model for the iron and zinc contents of 10 cowpea genotypes (G) evaluated in ten environments (E) of the Colombian Caribbean 2017-2018.

SV: source of variation; DF: degrees of freedom; SS: sums of squares; MS: mean squares; % SS: percentage of the sum of squares accumulated;  $PC_1$ ,  $PC_2$ , ...,  $PC_{10}$  = principal component 1, 2, ..., 10, respectively; \*\* = significant at 1%; \*: significant at 5%; ns: not significant.



**Figure 1**. Biplot of the AMMI model for the iron content (mg kg<sup>-1</sup>) of 10 cowpea genotypes evaluated in 10 environments of the Colombian Caribbean. The codes of environments and genotypes are described in Materials and Methods.

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More favorable environments for iron contents and above the average were observed in Montería (MO7B), Leticia (LE8A) and Polonuevo (PN7B), while the most unfavorable were in Sampués (SA8A) and Villanueva (VI8A). In Sampués (SA7B), Mahates (MA7B), Polonuevo (PN8A) and Cereté (CE7B and CE7A), slightly lower than average contents were recorded (Table 2). On the other hand, genotypes 1, 2, 3, 4, 5, and 8 had above-average iron content, while genotypes 6, 7, 9 and 10 were below. It is important to note that environments considered unfavorable can be improved through management, especially with adjustments to fertilizers with these micronutrients.

The dispersion of the environments was higher than that of the genotypes and much more contrasting due to macroenvironmental and microenvironmental, biotic and abiotic factors. Likewise, it is highlighted that genotypes 2 and 3 with scores close to zero in PC1 are considered very stable with respect to the evaluated environments and their iron contents are above the average (Figure 1).

Genotypes 1, 2, 3, 4, 5, 8 and 9 interacted positively with five of the 10 environments: LE8A,

MO7B, VI8A, SA8A and CE7B, and negatively with the remaining five, while genotypes 6, 7 and 10 interacted positively with the CE8A, MA7B, PN7B, PN8A, and SA7B environments, and negatively with the rest (Table 2). Positive interactions increase the estimated iron content with major effects and negative interactions decrease it. The Sampués environment score (SA8A) was the lowest, close to zero, so interaction with the 10 genotypes was also low, i.e. this environment does not significantly add or substract from the iron content of each genotype.

The Villanueva (VI8A) environment presented the highest positive score in the first axis of the PCA (Figure 1), with a strong influence on the positive interaction with genotypes 1, 4, 5, 8, and 9, and negative with genotypes 6, and 10, increasing and decreasing, respectively, the estimated iron content by adding the interaction effect to the main effects. In general, the environments with the lowest contribution to the GEI were Sampués (SA8A and SA7B), while the other environments contributed to a greater degree. Similar results, have been reported in pearl millet by Anuradha et al. (2017) and maize by Mallikarjuna et al. (2015).

	Genotype												
Environment	1	10	2	3	4	5	6	7	8	9	Mean	PC1	PC2
CE7B	55.0	57.3	56.0	62.4	62.6	59.3	53.1	49.7	62.4	58.5	57.6	0.55	-0.62
CE8A	61.6	58.4	58.5	57.1	59.0	61.0	53.4	63.9	61.2	57.1	59.1	-0.54	-0.45
LE8A	74.2	69.5	71.3	69.9	73.4	73.7	68.8	66.2	72.9	77.7	71.8	0.48	0.10
MA7B	58.5	55.4	61.6	59.0	57.4	57.9	64.3	58.0	57.8	59.1	58.9	-0.29	0.64
MO7B	70.0	58.1	67.8	69.2	71.7	72.4	72.5	62.7	72.6	71.5	68.9	0.55	0.68
PN7B	63.3	61.6	63.9	65.2	65.8	64.8	69.4	71.5	65.8	64.9	65.6	-0.82	0.43
PN8A	58.2	53.9	61.2	63.2	60.0	56.3	56.5	65.0	59.5	54.6	58.8	-0.80	-0.32
SA7B	58.0	54.6	57.8	55.2	55.0	57.1	60.1	54.4	59.0	52.9	56.4	-0.18	0.31
SA8A	46.6	46.1	56.3	49.8	53.4	49.5	45.1	47.3	57.4	45.8	49.7	0.05	-0.84
VI8A	60.8	46.7	58.6	55.9	55.5	56.3	51.9	43.1	58.2	55.3	54.2	1.00	0.06
Mean	60.6	56.2	61.3	60.7	61.4	60.8	59.5	58.2	62.7	59.7	60.1		
PC1	0.23	-0.18	0.06	0.05	0.19	0.23	-0.14	-1.00	0.23	0.34			
PC2	-0.13	0.08	0.48	0.17	0.20	-0.13	-0.66	0.00	0.39	-0.42			

**Table 2**. Iron content (mg kg<sup>-1</sup>) for 10 cowpea genotypes in 10 environments and means of the AMMI model with the first two principal components (PC1 and PC2) of the genotype-environment interaction analysis.

The code of environments and genotypes are described in Materials and Methods.

The closer a genotype is to the origin of the ordinate, the less contribution it adds to the effect of the genotype-environment interaction, be it positive or negative, and presents greater stability in the iron content. The most stable genotypes differ from the rest in the efficiency of iron use and absorption. According to Figure 1, the genotypes that contributed the least to the effect of the interaction

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were 2 and 3, although their iron contents were higher than the general average, so they could be recommended to farmers and genotype 7 showed greater interaction and consequently instability.

The AMMI2 biplot (Figure 2) has the first principal component (PC1) on the abscissa axis with 42.72% of the SS and the second principal component (PC2) on the ordinate axis with 22.21%, both totaling 64.93% of the sum of squares of the GEI.

In Figure 2 it is corroborated that genotypes 2 and 3 are closer to the origin of the coordinates, so they can be considered as those with the highest buffering capacity or stability. On the other hand, the Sampués (SA7B) and Mahates (MA7B) environments contributed the most to the stability of the genotypes, although their iron accumulations were below average. The remaining eight environments, the furthest from the origin in the biplot and outside the polygon, contributed more to the GEI. In addition, genotypes 6, 7, 8, 9 and 10, located at the vertices of the polygon, were the ones that responded most to environmental stimuli either positively or negatively, depending on the sign and magnitude of the scores.

In the orthogonal projections and closeness of the genotypes, and in the product of the PC1 scores for environments and genotypes, it was observed that genotypes 1, 4, 5, 8, and 9 interact positively and to a greater extent with environments VI8A, LE8A, MO7B and CE7B, while genotypes 6, 7 and 10 did so with PN7B, PN8A and CE8A, showing specific adaptability to these environments. The effect of the positive or negative interaction of genotypes 2 and 3 with all environments was almost zero, showing stability in the iron content.



**Figure 2**. Biplot with the principal components PC1 and PC2 of the AMMI model for the iron content (mg kg<sup>-1</sup>) of 10 cowpea genotypes evaluated in 10 environments of the Colombian Caribbean. The codes of environments and genotypes are described in Materials and Methods.

The biplot for zinc content is presented in the Figure 3. This representation contains 93.03% of the sum of squares of the AMMI model and 40.59% of the sum of squares of the GEI and constitutes the AMMI1, that is, the reduced model in which the GEI is explained only with the first axis of the PCA, as explained above. Zinc contents above the average were observed in the MO7B, LE8A, CE7B, PN7B, MA7B, SA7B and PN8A environments; with MO7B having a great advantage with respect to the other environments (Table 3).

Lower than average zinc contents were observed in CE8A, VI8A, and SA8A. On the other hand, genotypes 1, 2, 3, 4, 6 and 8 presented zinc

contents above the average, and specific adaptability to SA7B, LE8A, MO7B, PN7B and MA7B environments. Genotype 6 stands out for its increased zinc contents and higher positive score in LE8A, MO7B and SA7B environments. The rest of the genotypes contents lower zinc contents (Table 3). Among the most stable due to their lower contribution to interaction, are genotypes 1, 5 and 9, with genotype 1 being the one with the best overall adaptability, as it shows greater accumulation of zinc (Figure 3).

Genotypes 2, 3, 4, 7 and 8 interacted positively and to a greater extent with the PN8A, SA8A and VI8A environments, and negatively with the rest of the environments, while genotypes 6 and 10 interacted positively and with greater effect with the SA7B, LE8A and MO7B environments, as can

be verified by multiplying the scores of the environments and genotypes in the first axis of the PCA.



**Figure 3**. Biplot of the AMMI model for the zinc content (mg kg<sup>-1</sup>) of 10 cowpea genotypes evaluated in 10 environments of the Colombian Caribbean. The codes of environments and genotypes are described in Materials and Methods.

Environment	Genotype											
	1	10	2	3	4	5	6	7	8	9	Mean	PC1
CE7B	46.0	46.6	45.8	46.8	45.3	46.0	46.1	45.0	46.5	45.8	46.0	0.14
CE8A	37.4	37.6	36.6	36.8	36.6	37.4	38.5	38.3	36.8	35.8	37.2	-0.05
LE8A	44.5	47.7	43.7	43.8	44.8	43.6	49.5	42.9	44.0	43.8	44.8	-0.76
MA7B	45.3	42.1	43.5	46.3	44.3	44.0	45.4	41.6	43.0	43.0	43.9	0.00
MO7B	49.0	46.1	47.6	46.4	47.0	46.8	51.5	43.3	46.9	47.0	47.1	-0.56
PN7B	44.0	44.6	47.4	45.5	45.4	43.0	45.8	40.5	43.0	42.8	44.2	-0.05
PN8A	41.7	43.8	42.9	44.0	42.7	40.8	37.9	39.7	43.3	39.3	41.2	0.97
SA7B	43.5	41.9	39.9	42.3	42.1	42.7	45.8	38.2	41.6	40.6	41.9	-0.55
SA8A	30.9	29.5	29.5	31.6	34.8	31.8	31.2	31.7	34.7	29.8	31.6	0.55
VN8A	33.4	33.5	34.2	30.1	31.7	32.5	28.9	27.3	31.1	33.0	31.6	0.31
Mean	41.6	40.9	41.1	41.4	41.5	40.9	42.1	38.9	41.1	40.1	40.9	
PC1	-0.08	-0.29	0.23	0.27	0.25	0.06	-1.00	0.24	0.38	-0.06		

**Table 3**. Zinc content (mg kg<sup>-1</sup>) for 10 cowpea genotypes in 10 environments and means of the AMMI1 model with the first principal component (PC1) of the analysis of the genotype-environment interaction.

The code of environments and genotypes are described in Materials and Methods.

## **CONCLUSIONS**

The cowpea genotypes responsed differentially for the contents of iron and zinc in the grain, under the environmental condition of the Colombian Caribbean.

Cowpea genotypes 2 and 3 expressed greater stability and adaptability for iron contents, while genotype 1 demonstrated this for zinc; therefore, they can be cultivated in all the evaluated environments.

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