

Environmental Impact on Early Generations Selection and Its Implications on Breeding Strategies in Rice

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Abstract: Genotype-environment interaction poses an outmost barrier to the breeders in the process of development of improves rice varieties. Understanding the causes and extent of Genotype-environment interaction is highly useful to frame breeding objectives, identifying ideal test locations, and formulating varietal release recommendations. The objectives of this study were to: (i) know the effect of environment on gene action, (ii) assess yield stability across environment, (iii) know the ideal environment to discriminate the F3 population and (iv) assess ideal genotype to discriminate the environment. Thus, 120 rice families, 4 parents and local check were evaluated on 2 saline environments. Collected data were analyzed using SAS software version 9.2 and Genstat 18th. Non significant effect of environment on gene action was noticed for all the traits, tiller excepted. The genotype by environment interaction was highly significant implying differential performance of families at different locations. However, there was crossover $G \times E$ interaction, suggesting that different winners can be evaluated in different environments.

Keywords: genotypes and environment, interaction, implication, rice breeding, Niger.

1. INTRODUCTION

Rice (*Oryza sativa*) is the most important food crop that supplies over 21% of the calorific demands of the world population (Fasahat *et al.* 2014). In developing new rice varieties, the main trait that breeders consider is grain yield (Fitzgerald *et al.* 2008). However the environment affects the expression of quantitative traits such as grain yield, and different environments can affect genotypes differently. Phenotypic values are classically divided into genotypic (G), environmental (E) and genotype \times environmental interaction (G \times E) effects (Hallauer and Miranda Fo 1988). The environment is the sum of total physical, chemical and biological factors that influence the development of an organism (Nadarajan and Gunasekaran 2005). The GE interaction, is defined by Cooper and Byth (1996) as the variation in relative performance of genotypes in different environments. Genotype-environment interaction poses a major barrier to the breeder in the process of improving variety (Sangodele *et al.* 2013). This is, because it complicates the selection of superior genotypes, thereby reduces genetic progress (Romagosa and Fox 1993). Thus, if GE interactions are present, breeders need to identify stable genotypes with relatively consistent performance across a range of environments (Ouk *et al.* 2007).

Several methods have been used to quantify the GE interaction for yield. These include contrasts, Regression on Mean Model, Additive Main and Multiplicative Interaction (AMMI) Model, Principal Component Axis (PCA) Model (Yan 2011) and GGE biplot model. The latter is used in this study to work out genotypes by environment interaction. This is because it is simpler and more informative (Yan 2011).

The objectives of this study were to: (i) understand the effect of environment on gene action, (ii) assess yield stability across environments, (iii) identify the ideal genotypes and environment for yield.

2. MATERIAL AND METHODS

The study was carried out on two sites of Niger country namely, Saga with a latitude of 13°28'N and a longitude of 2°08' and Sekoukou latitude of 13°15'N and a longitude of 2°22'. The soils characteristics are presented in the Table1.

Table 1: Physical and chemical characteristic of the study sites soils

Sites	pH	SAR	CEC (meq/100g)	EC (dS/m)	Na/k (%)	Clay (%)	Silt (%)	Sand (%)
Saga	5.4	14.38	11.94	12.30	118.6	11.6	13.1	75.3
Sekoukou	5.2	12.46	18.96	8.6	60.8	48.4	26.6	25

SAR=sodium adsorption ration, CEC= cation exchange capacity, EC= electrical conductivity, Na/K = sodium and potassium ration

Two salt tolerant genotypes: IRR113 and NSIC RC106 (Souleymane *et al.* 2016) and two susceptible farmers varieties IR1526 and Gambiaka have been crossed in full diallel. The first generations F₁ from these crosses were bulked and selfed to get 12 F₂ families (Kol-2, Kol-15, Kol-14, Kol-11, Kol-5, Kol-4, Kol-31, Kol-29, Kol-27, Kol-25, Kol-23, Kol-21). The F₂ families were advanced to have F₃ families. A random sample of one hundred and twenty (120) F₃ families derived from F₂ individual plants were taken for evaluation in farmer's field affected by salt problem. In addition 4 parents and a farmer preferred variety (NERICA-L49) served as check were also evaluated in the same condition. The experimental design was 25*5 Alpha lattice with three replications. Each block was constituted with 25 lines with 10 plants on lines and each line was constituted of one F₃ family. The inter-plant space was 0.2 m and the between line space was 0.5 m. At the maturity plant height, total tillers per plant, reproductive tillers per plant, number of panicle, panicle weight, and paddy yield were scored. The time to 50% flowering, time to 85% maturity were also recorded (Souleymane *et al.* 2015). Data were analyzed using SAS software version 9.2. A general ANOVA was performed using SAS Glm procedure with random effect Model.

The Hayman diallel model (Hayman 1954) was used for gene action study. The model used is:

$Y = U + \text{rep} + a + b + c + d + a*\text{rep} + b*\text{rep} + c*\text{rep} + d*\text{rep}$, where

- U = grand mean; rep= replication effects; a = additive effects; b = dominance effects;

b is partitioned into: b1 that indicates direction of dominance (unidirectional if significant; equiv. to Parent vs. crosses contrast); b2, tests asymmetry of alleles; b3, shows that some dominance is peculiar to some crosses. c = additive maternal effects; d = maternal interaction effects;

- $a*\text{rep} + b*\text{rep} + c*\text{rep} + d*\text{rep}$ = interaction of the reps with the model components.

Genstat software version 18th was used for genotype and environmental interaction study. The GGE biplot methods were used for the analysis (Yan 2011).

3. RESULTS

The environments (sites) were significantly different for all the traits measured (Table 2). Highly significant differences also existed among F₃ families. Family's performances were highly significantly influenced by environment effects. This is so because families and environment interaction was highly significant for all the traits.

Table 2: families and environment interaction

Source	d.f.	Flw	Tnum	Pnum	height	Tpwt	Pwt	Gwt
Env	1	166821.84***	882752.6***	623944.43***	342190.56***	838150.7***	248.28***	342967.16***
Family	124	1098.36***	2812.42***	2742.03***	1098.36***	4321.86***	8.15***	2833.79***
Env*family	124	411.29***	2667.03***	2475.01***	411.29***	4434.37***	8.94***	3650.05***

***= very highly significant. Pwt= panicle weight, Tpwt = total panicle weight, Gwt = grain weight, Pnum = panicle number, Tnum= tiller number, and Flw= time to 50% flowering.

Effect of environment on gene action (additive, dominance, and maternal effect) existed but was not

significant for all the traits measured (Figure 3). However for tiller number, significant interaction existed between dominance effect and the environment.

Table 3: genes action and environment

Interaction	d.f.	Flw	Tnum	Pnum	height	Tpwt	Pwt	Gwt
Env*a	3	18.76ns	0.95ns	26.43ns	25.22ns	60.86ns	0.03ns	54.44ns
Env*b1	1	25.66ns	50.33ns	45.91ns	1.27ns	166.64ns	0.96ns	76.28ns
Env*b2	3	3.65ns	169.39*	107.97ns	13.06ns	163.89ns	0.61ns	91.47ns
Env*b3	2	3.32ns	4.93ns	4.23ns	6.62ns	6.37ns	0.01ns	0.27ns
Env*b	6	7.21ns	94.73ns	63.05ns	8.95ns	111.84ns	0.47ns	58.54ns
Env*c	3	8.00ns	18.35ns	35.39ns	0.25ns	192.60ns	0.28ns	130.65ns
Env*d	3	2.01ns	18.09ns	23.34ns	10.07ns	57.25ns	0.08ns	22.18ns

Pwt= panicle weight, Tpwt = total panicle weight, Gwt = grain weight, Pnum = panicle number, Tnum= tiller number, and Flw= time to50% flowering.

The yield scatter plot (Figure 1) showed that the two sites were different and formed two distinct mega-environments. The angle between the two environment is slightly larger than 90°, implying that the genotype/environment interaction is moderately large. The environment one has a vector longer than environment two. This implies that the environment 1 discriminate more efficiently the families than environment 2. All the families were distant from the origine this means that they were highly responsive to the environment effect. The families which won in the differents sectors were 95, 114, 96, 17, 121 and 118. The families that performed well in environment 1 were 101, 95, 88, 92, 114, etc.... The families which won in the environment 2 were: 17, 96, 107, 22, 121, etc...

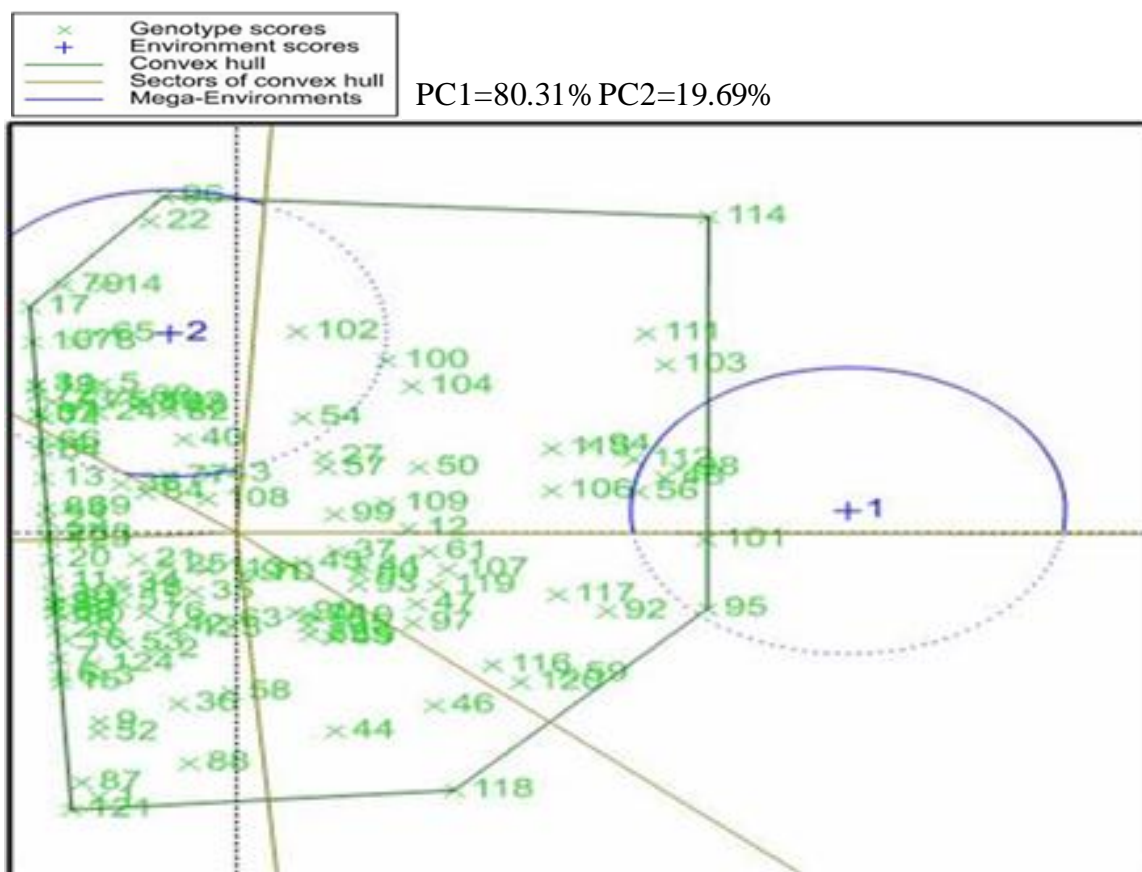


Figure 1: GGE biplot identification of winning genotypes and their related mega-environments

The ranking plot (Figure 2) shows a point (0) that is the average of all the environments and a line from the origin to the mean environment. Any families above the origin perform higher. The best families were: 114, 111, 103, 112, 84, 11, etc..... The most high yielding and stable families were: 111, 103. However, the family 114 that was the most performing in term of yield was more responsive to the environment.

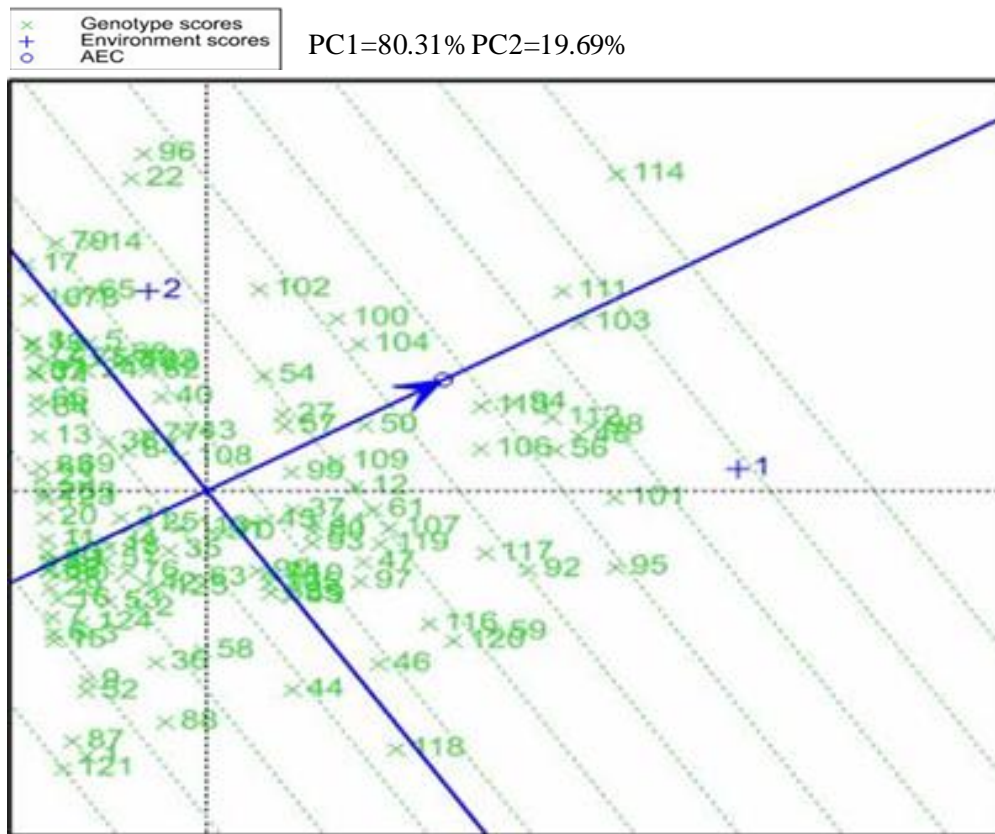


Figure 2: Stability of rice genotypes for yield

The genotypes comparison plot (Figure 3) shows the ideal genotype in the center. The family closer to the ideal is the best to discriminate the environments. Thus, family 114 is the ideal followed by 111, 103 etc....

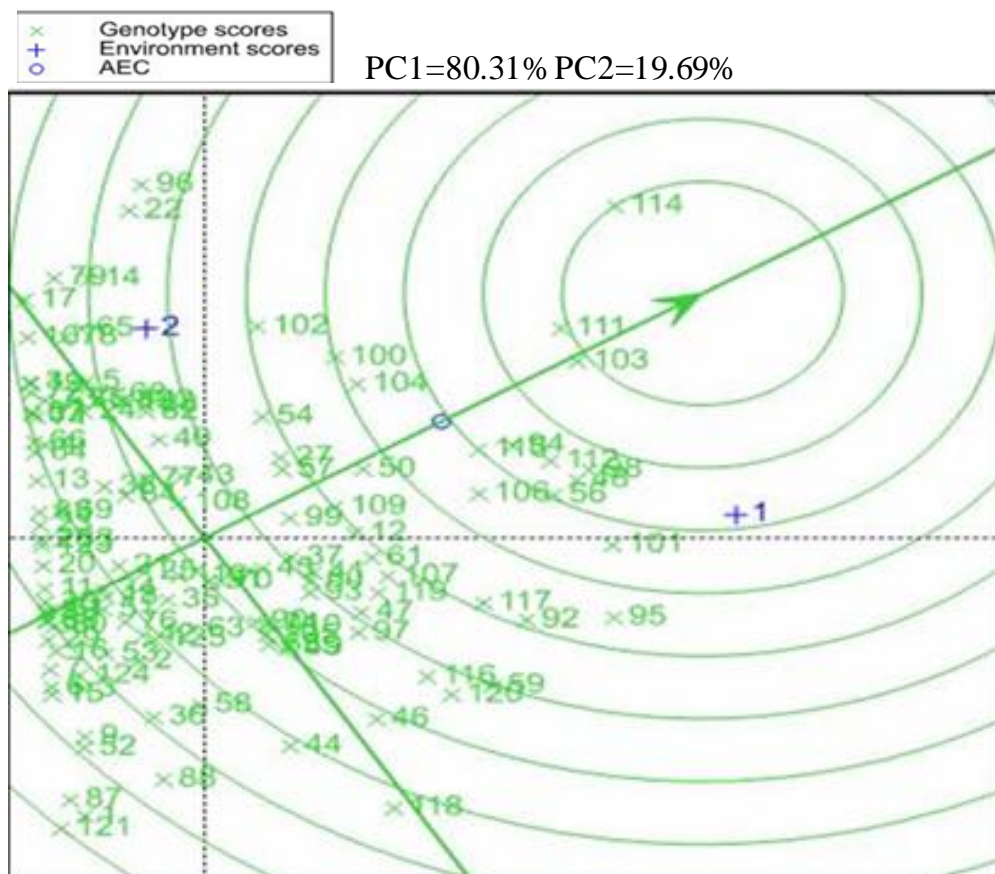


Figure 3: Identification of best families based on ideal family

NB: AEC, average environment coordination

The environment comparison plot (Figure 4) shows that the best environment to discriminate the families was the environment 1. This is because it is the closer to the ideal environment.

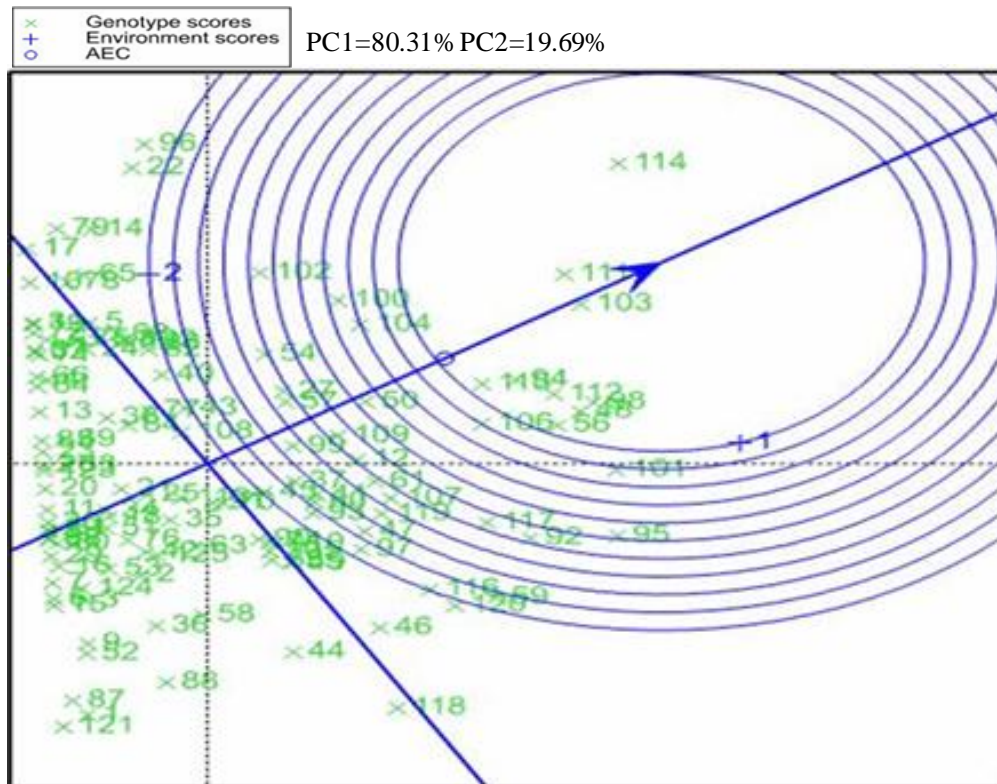


Figure 4: Identification of best environments based on ideal test environment

NB: **AEC** = average environment coordination

4. DISCUSSION

Understanding the causes and extent of $G \times E$ interaction is highly useful to frame breeding objectives, identifying ideal test locations, and formulating varietal release recommendations (Krishnamurthy *et al.* 2017). It is also of paramount importance to evaluate the adaptability and stability of the genotypes.

Results showed significant differences among location. This implied environments prevailing at the two sites were dissimilar. The genotype by environment interaction was highly significant for all the traits. This implied differential performance of families at different locations (Simmonds Smartt, J.). This may be due to the variability salt stress level across environments and within the same environment. Significant differences of environments increase substantially genotype by environment interaction and can reduce selection gain (Betran *et al.* 2004; Cooper *et al.* 2006). According to Bose *et al.* (2012) highly significant genotype \times environment ($G \times E$) interaction suggests that the genotypes interacted considerably with environmental conditions. Thus, significant genotype and environment necessitates more testing over larger number of locations (multi-environment trials) to evaluate genotype adaptation of fixed lines in the future (Cooper *et al.* 1999). Hence, according to Grando and Ceccarelli (2009) salinity is unpredictable and variable. There was crossover $G \times E$ interaction: rank of the genotypes changes in different environments, suggesting that different winners can be evaluated in different environments (Braun *et al.* 1996; Troyer 1996; Yan and Hunt 2000; Yan and Tinker 2006). There was a location (environment 1) that best discriminates the genotypes. That environment (Saga) should be used to improve the selection efficiency.

According to Singh *et al.* (2010) genotypes with a high yield mean, under stress environments are the most suitable, stable and adaptable for sustainable productivity in problem soils. The result results showed that the most high yielding and stable families were: 111, 103. These families should be advanced for lines fixation and release to farmers.

According to Cooper *et al.* (1999) $G \times E$ interaction acts to reduce heritability. Thus, where $G \times E$ interactions exist, a condition for any estimate of heritability to be reliable is that the results must be based on a sample of environments that are representative (Cooper *et al.* 2006).

5. CONCLUSION

Rice lines performances were highly and significantly influenced by environments effects. The environments also affected differently lines yields implying environments prevailing at the two sites were dissimilar. There was crossover $G \times E$ interaction suggesting that different winners can be evaluated in different environments. The environment one (Saga) has a vector longer than environment two. Thus, is discriminate more efficiently the families than environment2 (Sekoukou). The most high yielding and stable families were: 111, 103. However the family 114 that was the most performing in term of yield was more responsive to the environment effect. However, family 114 is the ideal followed by 111, 103 etc.... The best environment to discriminate the families was the environment 1.

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