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Article

Genetic Variation and Heritability for Hydrogen Cyanide in Fresh Cassava Roots

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Abstract: Breeding for low hydrogen cyanide (HCN) varieties is a major objective of programmes targeting boiled cassava food product. To enhance breeding for low HCN, knowledge on genetic variation and trait heritability is essential. In this study, 64 cassava clones established across four locations and evaluated for HCN using three HCN assessment methods; 1 to 9 scale, 0 ppm to 800 ppm scale and a quantitative assay based on spectrophotometer readings (HCN_Spec). Data were also collected on the weather variables precipitation, relative humidity, and temperature. Significant differences were observed among clones ($P < 0.001$) and locations ($p < 0.001$). There were also significant ($p < 0.001$) clone by environment interactions. Locations Arua and Serere posted higher HCN scores among clones and were associated with significantly higher ($p < 0.001$) mean daily temperature (K) and lower relative humidity (%) across 12-hour and 18-hour intervals. Within locations HCN broad sense heritability estimates ranged 0.22 – 0.64 while combined location heritability estimates ranged 0.14 to 0.32. Relationships between the methods were positive and strong ($R^2 = 0.75 – 0.92$). It is expected that the information herein will accelerate efforts towards breeding for low HCN cassava varieties.

Keywords: breeding efficiency, end user preference; phenotyping; selection; genotype by environment interaction

1. Introduction

Cassava (*Manihot esculenta* Crunz) is a staple for more than 600 million people in Eastern, Western and Southern Africa (1). These regions produce and consume more than 54% of the world's cassava, underpinning the importance of the crop to their livelihoods. The crop's clonal nature, ability to survive in marginal soils, tolerance to drought (2) and perennial nature which allows for peace meal harvest allow for a diversity of uses for food, feed and industry, making it an important food security crop (3,4,5). However, cassava's usefulness is limited by presence of the cyanogenic glucosides, namely linamarin and lotaustralin in the leaves and roots (6,7). Linamarine readily hydrolyzes to glucose and acetone cyanohydrin in presence of *linamarase* enzyme and in neutral or alkaline medium, acetone cyanohydrin decomposes to liberate cyanide and a hydrogen ion (8).

Production of hydrogen cyanide (HCN) in plants is said to be a defense mechanism against herbivores and pathogens (9,10). Total cyanogenic concentration in cassava may vary with cultivar, environmental conditions, cultural practices, and plant age. Indeed, HCN content in the range of 2 ppm to >1000 ppm has been reported (8,11). Cassava is thus classified as "sweet" if it has root HCN content < 100 ppm on wet basis and "bitter" if it has root HCN > 100 ppm (12,13). In fact, cassava with HCN content > 10 ppm on dry weight basis or > 100 ppm on fresh weight basis is considered not safe for human consumption and should undergo further processing before consumption (14).

Dietary intake of high amounts of HCN can be toxic to and/or lethal (15,16). Prolonged dietary exposure to HCN is linked to degenerative conditions of the nervous system; konzo and tropical

ataxic neuropathy (17–19). Despite these risks, some communities still cultivate bitter (high HCN) varieties because they are not stolen by human or rodents (20). The rate of adoption of new crop varieties is dependent upon end user preferences (21-24).

Consumers of boiled cassava have preference for non-bitter cassava varieties (3,25,26), characterized by low HCN content. HCN content is thus a major component of cassava root quality for boiled or steamed cassava roots, a product that is desired by consumers of cassava. In the Amazon region, farmers readily classify cassava varieties into sweet, i.e. aipim, macaxera or table varieties and bitter varieties which are said to be for industry (27), while in Malawi, farmers ably classify their varieties as either “cool” or “bitter” depending on HCN levels (28). Owing to the increasing interest in boiled cassava, efforts are underway to initiate breeding for low HCN cassava varieties. However, the success of these efforts largely depends on amount of variation present and the proportion of this variation that is heritable (28). In addition to genotypic differences, variation in fresh cassava root HCN content is said to be influenced by the soil nutrient status (30) and climatic factors such as precipitation and temperature (31).

Phenotyping for HCN content of fresh cassava roots by cassava breeding programs is commonly done using any of three variations of the picrate method. These include 0 ppm to 800 ppm color scale (8), the modified picrate method (6,32) which allows for collection of semi-quantitative data using a spectrophotometer, or another modification of the picrate method suggested by (34) which follows a 1 to 9 color scale. All these methods have ontologies for data handling in the cassava breeding database; Cassavabase (35) and are thus deployed by major cassava breeding programs. Whether or not these methods provide comparable results is unknown. Furthermore, it is not known whether sufficient genetic variation exists in available germplasm to initiate long-term breeding for low HCN. The effect weather variables on HCN phenotypic expression variability within the available germplasm also remains unknown. Thus, in this study, we sought to (a) access the extent of genetic variation and heritability for HCN within the study germplasm (b) compare accuracy of three HCN phenotyping protocols (c) access the relationship between the weather variables; precipitation, temperature, and relative humidity and fresh cassava root HCN phenotypic expression.

2. Results

2.1. Variability for HCN in fresh cassava roots

Highly significant differences among clones ($P < 0.001$) and locations ($P < 0.001$) were observed for all three HCN assessment methods. Furthermore, there was significant clone by environment interactions for the methods (Table 1).

Table 1. Analysis of variance for fresh cassava root HCN content across locations.

Source of variation	Degrees of freedom	Mean Square		
		1-9_Scale	HCN_Spec	HCN_0-800_Scale
Clone	63	5.069***	57706***	130338***
Location	3	127.548***	1127633***	3384185***
Replication	1	1.133	670202***	33359
Accession: Location	149	1.571*	27349***	33731
Residuals	147	1.085	14280	27283

Significance codes: 0.0001 '***' 0.001 '**' 0.01 '*' 0.05 '.' 1-9_Scale = HCN scored on 1 to 9 categorical scale, HCN_Spec = HCN values in parts per million computed from spectrophotometric readings, HCN_0-800_Scale = HCN scored on a 0 to 800 parts per million categorical scale.

On the 1-9_Scale, fresh root HCN content ranged from 2 to 9 with a combined location average of 6.56. The highest average HCN content was recorded in Arua (7.70) followed by Serere (7.58), while Namulonge (5.45) had the lowest average HCN content (Table 2). On the 0 to 800 scale, fresh root HCN content ranged from 65 ppm to 800 ppm with a combined location average of 430.21 ppm. The

highest average HCN content (614.24 ppm) was again recorded in Arua, while the least (236.03 ppm) was again recorded in Namulonge (Table 2). For the HCN_Spec method, fresh root HCN content ranged from 119.6 ppm to 676.3ppm with a combined location average of 504.1 ppm. However, the highest average fresh root HCN content (608.53 ppm) was recorded in Serere while the least (399.39 ppm) was recorded in Tororo. Generally, HCN observations at Namulonge were comparable to Tororo while observations for Serere were comparable to Arua (Table 2).

Within sites, broad sense heritability ranged 0.46 to 0.54 on the 1 to 9 scale, 0.24 to 0.64 on the 0 to 800 ppm scale and 0.22 to 0.56 using the spectrophotometric method. Across locations, broad sense heritability was 0.14 for the spectrophotometric method, 0.26 for the 1 to 9 scale and 0.32 for the 0 to 800 ppm scale. The genetic component of the variation (V_g) was 0.48 for the 1 to 9 scale, 3596 for the spectrophotometric method and 13990 for the 0 to 800 scale. (Table 2).

Table 2. Summary of phenotypic variation, variance components and broad sense heritability for HCN in fresh cassava roots across locations and HCN phenotyping methods.

Location	HCN_1-9_Scale	HCN_0-800_Scale	HCN_Spec
Arua	7.70 ^a (0.46)	614.24 ^a (0.64)	601.85 ^a (0.50)
Serere	7.58 ^a (0.54)	595.68 ^a (0.24)	608.19 ^a (0.22)
Namulonge	5.45 ^b (0.50)	277.48 ^b (0.52)	407.0 ^b (0.56)
Tororo	5.75 ^b (0.48)	263.03 ^b (0.56)	399.39 ^b (0.40)
CV (%)	15.6	38.5	24.6
V_g	0.48	13990	3596
V_{ge}	0.35	4943	7748
V_{Res}	1.04	24897	14093
H^2	0.26	0.32	0.14

Means followed by the same letter are not significantly different at 5 % alpha level. HCN = Hydrogen Cyanide, Spec = Spectrophotometer, CV = Coefficient of variation, % = Percentage, V_g = Genotypic variance, V_{ge} = Genotype-by-environment variance, V_{Res} = Residual Variance, H^2 = Broad sense heritability. Broad sense heritability within locations is indicated in brackets for each method and location.

On the 1 – 9_Scale, Arua had clones scoring 4 to 9 with a high median value of 8. The distribution was skewed to the right (Figure 1). In Serere scores ranged 3 to 9, with 50% of the clones scoring between 7 and 8, there by skewing the distribution to the right. In Tororo, HCN scores ranged 3 to 9 with 50% clones scoring between 5 and 6 while in Namulonge scores ranged 2 to 9, with a median of 6 and a near uniform distribution of scores across the scale (Figure 1). However, for the 0 – 800_Scale, HCN scores ranged 100 to 800 ppm in Arua with 75% clones scoring 400 – 800 ppm, there-by skewing the distribution to the right. Similarly, in Serere, HCN scores ranged 100 to 800 ppm, with a median score slightly below 700 ppm. The distribution of scores was skewed to the right. In Namulonge and Tororo, the median HCN scores were 200 ppm and 150 ppm respectively with scores ranging from less than 50 ppm to 800 ppm. For both locations, the distribution was skewed to the left (Figure 1). A similar trend was observed using the spectrophotometric method. Generally, Arua and Serere had high HCN values, thereby skewing the distributions to the right, while Namulonge and Tororo had uniform distributions (Figure 1).

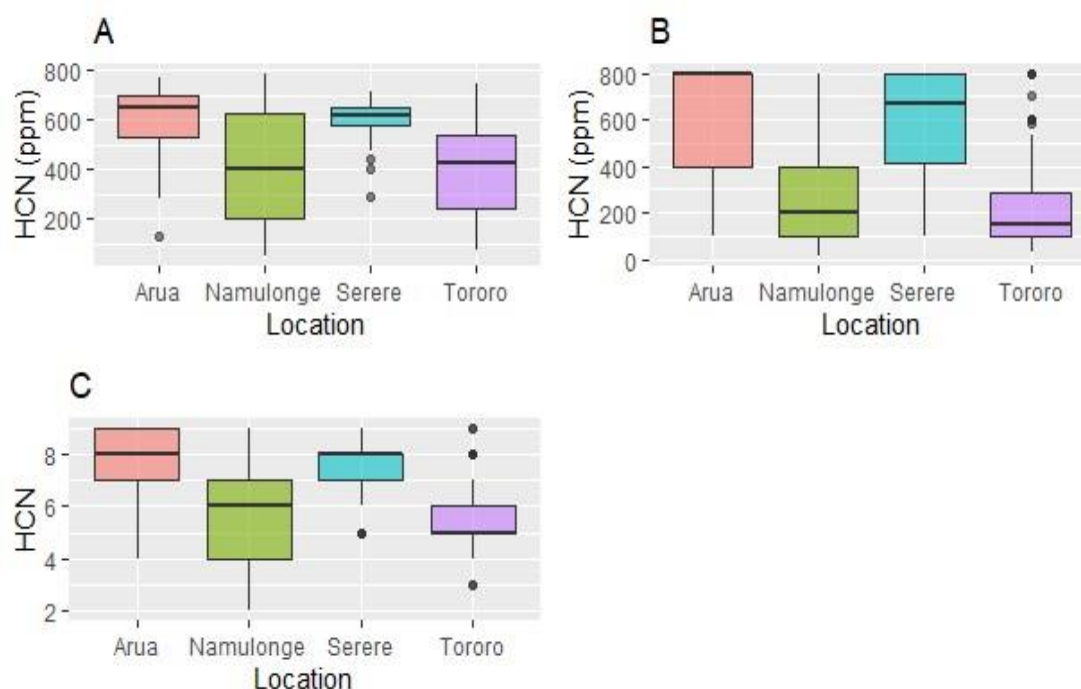


Figure 1: Variation of fresh cassava root HCN content across test sites based on three assessment methods: the spectrophotometric method (A), the 0 to 800 ppm categorical scale (B) and the 1 to 9 categorical scale (C).

2.2. Accession ranking for fresh cassava root HCN content by BLUPS

BLUPS consistently ranked four of the five bitter local farmers' clones among the top 10 for high HCN content (Table 3). Over all, six clones; Tongolo, Nyamatia, Kazimwenge, Quinine, UG16F303P006 and UG16F158P004 were consistently ranked among the top 10 for high HCN content. On the other hand, 5 clones; UG16F290P295, UG16F293P169, UG16F057P001, UG16F001P013 and UG16F290P075 were consistently ranked among the top ten for low HCN content. Surprisingly, the local bitter farmers' accession Edwarat was ranked among the top 10 low HCN clones by both the 1 to 9 scale and spectrophotometric methods. Overall, 5 clones were consistently selected by all methods, 6 clones by 1 to 9 scale and 0 to 800 scale and 9 clones by 1 to 9 scale and spectrophotometric method (Table 3).

Table 3. Ranking of clones from highest to lowest HCN content using BLUP values for the three HCN evaluation methods.

1 - 9 Scale		0 - 800 Scale		HCN_Spec	
Accession	BLUP	Accession	BLUP	Accession	BLUP
TONGOLO	1.306	TONGOLO	257.776	KAZIMWENGE	63.186
NYAMATIA	1.085	NYAMATIA	225.384	UG16F300P068	59.25
KAZIMWENGE	0.883	QUININE	181.413	TONGOLO	56.682
QUININE	0.81	KAZIMWENGE	177.962	UG16F158P004	55.176
UG16F303P006	0.785	UG16F303P006	124.465	UG16F063P006	54.696
UG16F290P332	0.768	UG16F063P006	120.565	NYAMATIA	53.768
UG16F303P009	0.758	UG16F303P009	114.728	UG16F294P011	44.224
UG16F158P004	0.734	UG16F158P004	103.961	UG16F293P151	42.927
UG16F294P011	0.707	UG16F300P068	102.232	UG16F077P002	42.798
UG16F063P006	0.672	UG16F159P001	91.997	QUININE	36.43
UG16F300P068	0.669	UG16F043P008	83.378	UG16F303P009	35.75
UG16F318P051	0.609	UG16F318P051	82.859	UG16F290P047	33.975

UG16F159P001	0.572	UG16F290P332	80.568	UG16F043P008	33.19
UG16F318P035	0.552	UG16F294P011	69.998	UG16F290P332	30.111
UG16F043P008	0.545	UG16F318P035	67.153	UG16F159P001	29.036
UG16F293P050	0.462	UG16F323P082	61.891	UG16F323P082	28.527
UG16F077P002	0.348	UG16F302P027	61.351	UG16F318P035	26.041
UG16F293P069	0.344	UG16F315P053	60.3	UG16F318P053	25.734
UG16F290P047	0.34	UG16F290P047	47.06	NAMIKONGA2	25.507
UG16F318P053	0.262	UG16F293P069	46.438	UG16F322P033	22.637
UG16F323P082	0.257	UG16F214P012	44.873	UG16F318P051	22.541
UG16F322P033	0.255	UG16F308P058	41.801	UG16F020P002	22.485
UG16F302P027	0.236	UG16F322P033	37.727	UG16F293P069	22.004
UG16F041P009	0.235	UG16F292P017	36.901	UG16F290P293	21.505
UG16F046P005	0.202	UG16F293P050	23.988	UG16F303P006	20.572
UG16F292P017	0.192	UG16F077P002	23.843	IITA-TMS-IBA30572	18.846
IITA-TMS-IBA30572	0.16	UG16F041P009	21.471	UG16F041P009	16.237
UG16F293P151	0.153	UG16F318P053	12.605	UG16F293P050	13.71
UG16F290P042-1	0.143	NASEKE	11.08	UG16F290P215	12.73
UG16F315P053	0.125	NAMIKONGA2	10.964	UG16F290P042-1	6.772
UG16F214P012	0.122	UG16F150P002	4.425	UG16F046P005	3.467
UG16F150P002	0.089	UG16F290P042-1	2.417	UG16F292P017	1.369
UG16F308P058	0.053	UG16F293P151	-1.894	NASEKE	-0.392
NAMIKONGA2	0.045	UG16F290P293	-5.441	UG16F302P027	-0.439
NASEKE	0.042	UG16F046P005	-6.918	UG120156	-2.744
UG16F020P002	0.018	UG16F290P198	-9.205	UG16F290P049	-5.587
UG16F290P293	-0.008	UG16F009P019	-9.848	UG16F315P053	-5.714
UG16F290P198	-0.032	UG16F020P002	-23.228	UG16F214P012	-8.421
UG120156	-0.058	UG120156	-26.357	UG16F323P145	-8.687
UG16F0290P002-1	-0.161	IITA-TMS-IBA30572	-32.584	UG110017	-8.816
UG16F290P215	-0.181	UG16F225P003	-36.806	UG16F009P019	-8.835
UG16F290P049	-0.19	UG16F290P031	-42.968	UG16F290P198	-11.579
UG16F214P004	-0.221	UG16F0290P002-1	-43.338	UG16F0290P002-1	-12.556
UG16F009P019	-0.226	UG16F214P004	-43.419	UG16F290P031	-13.334
UG110017	-0.27	EDWARAT	-44.133	UG16F316P022	-16.707
UG16F290P031	-0.286	UG16F323P145	-53.321	UG16F308P058	-18.584
UG16F290P040	-0.337	UG16F290P006	-71.167	UG16F150P002	-19.392
UG16F323P145	-0.383	UG16F290P215	-77.821	UG16F314P005	-19.915
UG16F290P006	-0.395	UG16F290P128	-83.577	NAMIKONGA1	-20.388
UG16F314P005	-0.416	UG16F316P022	-86.441	UG16F214P004	-23.317
UG16F316P022	-0.419	NAMIKONGA1	-92.071	UG16F290P040	-25.523
UG16F290P073	-0.603	UG16F290P049	-97.144	UG16F225P003	-29.377
NAMIKONGA1	-0.613	UG16F293P082	-97.586	UG16F290P006	-33.47
UG16F225P003	-0.625	UG16F293P066	-101.141	UG16F303P005	-42.991
EDWARAT	-0.628	UG16F290P073	-102.508	EDWARAT	-51.843
UG16F290P128	-0.743	UG16F290P040	-107.26	UG16F290P073	-52.257
UG16F293P082	-0.753	UG16F314P005	-107.724	UG16F290P128	-52.338
UG16F303P005	-0.779	UG110017	-109.308	UG16F293P066	-52.881
UG16F290P295	-0.882	UG16F303P005	-119.431	UG16F293P169	-54.483
UG16F293P169	-1.039	UG16F057P001	-122.378	UG16F293P082	-56.428
UG16F057P001	-1.09	UG16F290P295	-125.102	UG16F057P001	-70.226
UG16F001P013	-1.262	UG16F293P169	-137.138	UG16F001P013	-71.706
UG16F293P066	-1.361	UG16F001P013	-189.676	UG16F290P295	-81.562
UG16F290P075	-1.574	UG16F290P075	-226.642	UG16F290P075	-101.391

Hydrogen Cyanide, Spec = Spectrophotometer, BLUP = Best Linear Unbiased Predictor.

2.3. Comparison of phenotyping method performance

2.3.1. Correlation between the methods

Regression analysis revealed a strong positive relationship between all the methods. The highest coefficient of determination ($R^2 = 0.92$, $p < 0.001$) was observed between the between the 1 to 9 scale and the 0 to 800 ppm scale (Figure 2A). Furthermore, there was a strong positive relationship ($R^2 = 0.85$, $p < 0.001$) between the 1 to 9 scale and the spectrophotometric method (Figure 2B) and between the 0 to 800 ppm scale and the spectrophotometric method ($R^2 = 0.75$, $p < 0.001$) (Figure 2C).

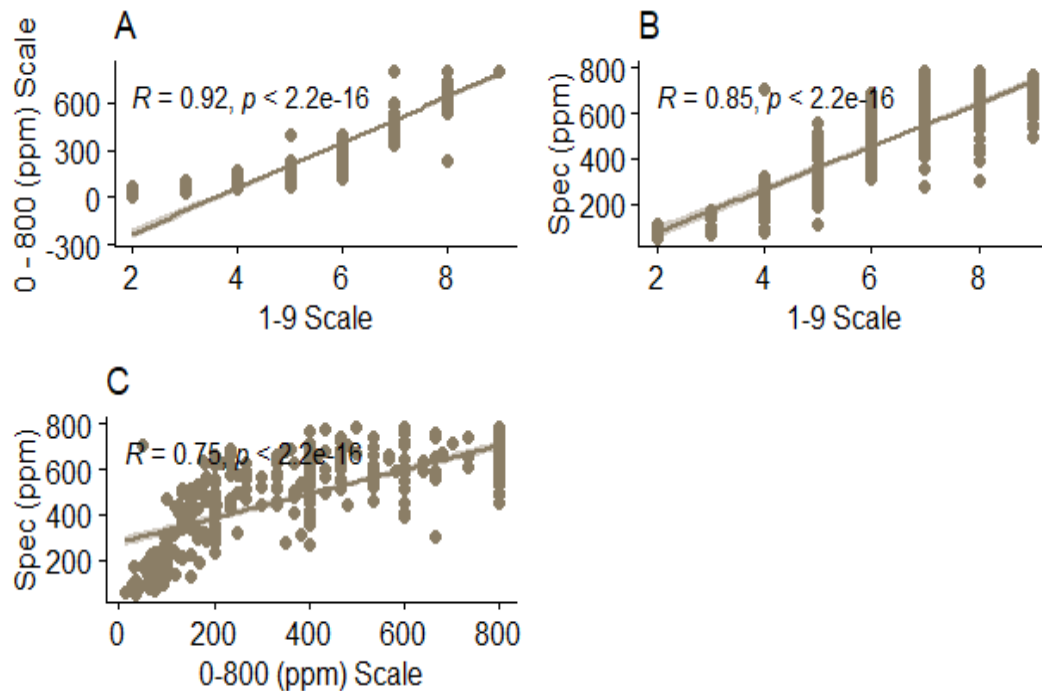


Figure 2. Scatter plots with regression lines and confidence intervals, depicting the relationship between the HCN phenotyping methods based on the picrate protocol. The relationship between spectrophotometric method and the 0 to 800 scale is shown in plot A. Plot B shows the relationship between the 0 to 800 scale and 1 to 9 scale while plot C shows the relationship between the spectrophotometric method and the 1 to 9 scale (C). ppm = parts per million, R = coefficient of determination, Spec = Spectrophotometric method.

2.3.2. Comparison of residuals

Large residuals were observed between 300 ppm and 600 ppm for both the 0 to 800 ppm scale and between 200 ppm and 400 ppm for the spectrophotometric method (Figure 3B). Residuals were more clustered towards the zero line for higher HCN readings (more than 400 ppm) for the spectrophotometric method while for the 0 to 800 ppm scale, this pattern was observed for lower HCN scores (less than 200 ppm). There was a random distribution of residuals for the 1 to 9 scale (Figure 3C).

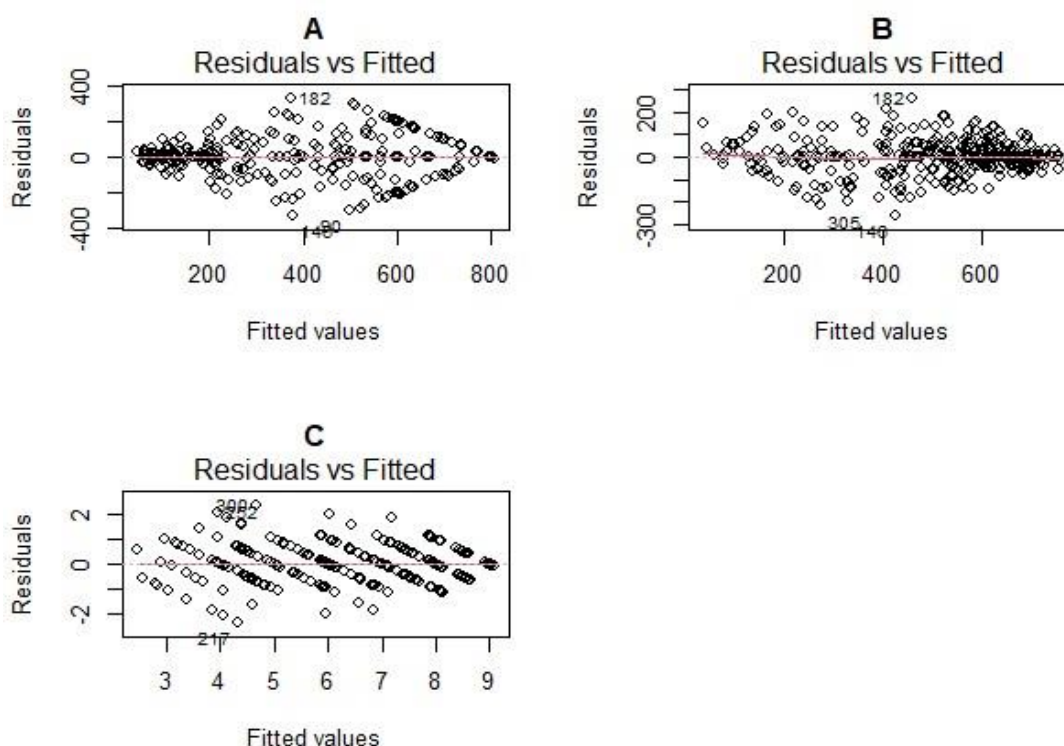


Figure 3. Residual vs Fitted Plots for the 0 to 800 scale (A), Spectrophotometric method (B), and 1 to 9 scale (C).

For a uniform comparison of residuals for all three methods, we computed the relative standard error (RSE). For the 1 to 9 scale, RSE was 17.4%, RSE was 25.7% for the Spec method and 41.32% for the 0 to 800 scale method, indicating that the 1 to 9 scale was the most accurate, followed by the spectrophotometric method, with the 0 to 800 scale being least accurate.

2.4. Variation in weather variables at experimental test locations

We compared locations for differences in precipitation (mm/day), average daily temperature (K) and relative humidity (%) at 12 and 18 hour intervals. There was significant variation ($p < 0.001$) among locations for 24-hour temperature, relative humidity at 12 hours and relative humidity at 18 hours (Figure 4). Except for precipitation, Namulonge was significantly different from all other locations. For precipitation, Namulonge received significantly higher precipitation ($p = 0.034$) than Serere but was not significantly different from Tororo and Arua (Figure 4A). Namulonge had the least average 24-hour temperature (20.97 K) followed by Tororo (22.16 K), Serere (22.99 K) and Arua (23.90 K) respectively. Arua was significantly hotter than Tororo ($p = 0.0003$) but did not differ from Serere (Figure 4B). For relative humidity at 12 hours, Namulonge was high (average = 70%) and significantly different from all other locations. However, there were no significant differences between the other locations (Figure 4C). For relative humidity at 18 hours, Namulonge had the highest relative humidity (77.03%) and was significantly different from other locations. This was followed by Tororo (66.17%) and Serere (60.98%). Arua with the least relative humidity (54.53%) was not significantly different from Serere. Serere was also not significantly different from Tororo but Tororo was significantly different from Arua ($p = 0.006$) (Figure 4D).

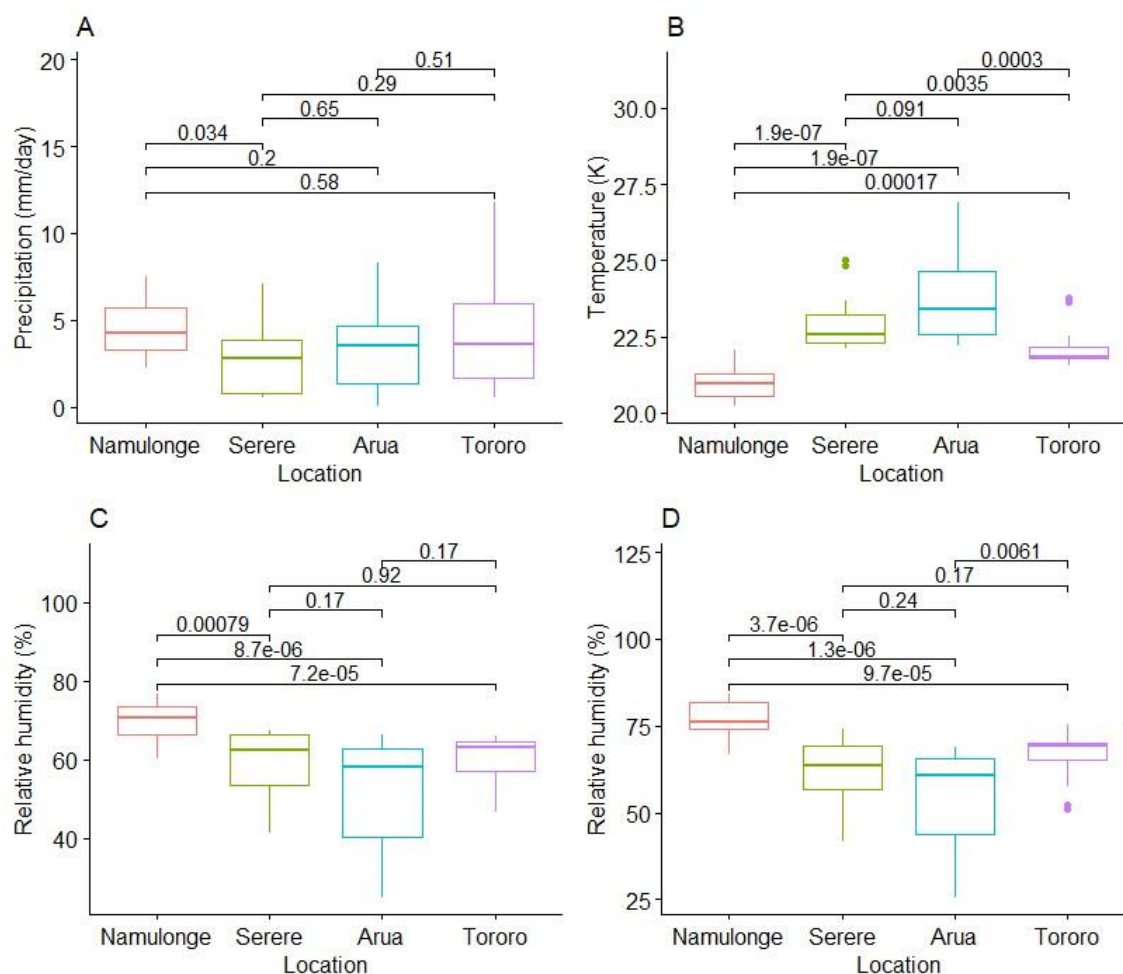


Figure 4. Variation of the weather variables; Precipitation (A), Temperature (B), Relative humidity at 12 hours (C) and Relative humidity at 18 hours (D) among test locations. The Kruskal-Wallis test was used to show significance or no significance in observed differences of weather parameters at $\alpha = 0.05$ across locations. % = percentage, K = Kelvin, mm/day = millimeters per day.

3. Discussion

3.1. Genetic variability and heritability for HCN

A major objective of this study was to establish whether significant genetic variation exists for HCN in available germplasm, and whether different assessment methods vary in precision. Highly significant variation for HCN among clones was observed (Table 1). Fresh root HCN content ranged 2.5 to 9 (mean 6.62) on the 1 to 9 scale, 165.2 ppm to 753.38 ppm on the 0 ppm to 800 ppm scale, and 119.50 ppm to 611.01 ppm on the HCN_Spec (Table 1). It is thus evident that there was variability for HCN within the evaluated germplasm that comprised both local and elite landraces. This favors selection for low HCN. The ranges of HCN reported in this study are comparable to those reported in earlier studies (36–38). Based on the 1 to 9 scale, the genetic component of the variation (V_g) was 0.48, which is higher than (V_g) = 0.21 reported in African germplasm from the International Institute for Tropical Agriculture (IITA), Nigeria and much lower than in Colombian (V_g) = 1.58 and Brazilian germplasm (V_g) = 2.59 (Ogbonna et al. 2020). The higher genetic variation for HCN in Ugandan germplasm compared to IITA's can be attributed to the deliberate inclusion of bitter landraces in the study i.e., Tongolo, Quinine, Nyamatia, Kazimwenge and Edwarat material whereas the high genetic variance reported for Brazilian and Colombian germplasm is explained by the fact that Latin America, particularly Brazil is the center of diversity of cassava. Given the large genetic variability for HCN reported for Latin American cassava clones, African cassava breeding programs targeting

low HCN product profiles can enhance the diversity in their populations through germplasm exchange with their Latin American counterparts. Germplasm exchange could also be beneficial for other traits like fresh root yield, where some African cassava breeding programs are not making appreciable genetic gains (39).

A clear understanding of trait heritability informs strategies to drive genetic gains (40,41) as heritability estimates are critical to selection decisions (42). Broad sense heritability (H^2) estimates were low to moderate across locations, $H^2 = (0.14 - 0.32)$. This could be explained by the large effect of environmental factors on HCN phenotypic expression. Furthermore, for all phenotyping methods, the residual variance was high, suggesting that better heritability estimates can be obtained with further optimization of the methods. These H^2 estimates are however comparable to estimates previously reported for West African germplasm ($H^2 = 0.27$) (38) and ($H^2 = 0.18$) (37) but much lower than ($H^2 = 0.76 - 0.82$) reported for Brazilian germplasm by the same authors using the 1 to 9 scale.

The diversity for bitter clones is maintained within African populations by farmers who selectively grow bitter varieties for various reasons (20,28,43). In Ugandan communities where communal livestock grazing is done, it is common practice for farmers to plant guard rows of bitter cassava landraces around their fields of the sweet cassava crop to "protect them from damage by livestock."

Whereas (37) suggested 3 categorizations of HCN on the 1 to 9 scale; sweet (1- 4), intermediate (4.1-5) and bitter (5.1-9), these categories are contestable given that the levels of taste were neither tested by a trained sensory panel nor validated by a consumer panel. Thus, there is no evidence to suggest that a genotype with a score of 5.1 or 6.0 will be perceived as bitter by the consumer. Rather, we suggest 4 categories that could be of practical implications for cassava breeding; low HCN (1 - 4.0) corresponding to 0 - 100 ppm, intermediate HCN (4.1 - 5.9) corresponding to 101 - 200 ppm, high (6.0 - 7.0) corresponding to (201 - 400 ppm) and very high (7.1 - 9.0) corresponding to 401 - 900 ppm.

Thus, for breeding efforts targeting boiled products (associated with minimal processing), an average score of 1- 4 is desirable, but less than 6.0 is acceptable. For the fresh-root chewing product profile (common in Tanzania) which is associated with no processing at all, the cut-off for selection should be at 4.0. This categorization is somewhat in alignment with (44) who recommended a maximum residue limit (MRL) of 200 ppm for cyanogenic glycosides in cassava for food. For product profiles involving extensive processing like drying, milling and fermentation, the cut off may be relaxed to include clones averaging 6.0 - 7.0 given that significant amounts of cyanogens are lost during processing (45-47). However, when initial fresh root HCN concentrations are very high, traditional processing methods may not be sufficient to detoxify the roots (45), resulting in fatalities (15, 16, 48). Depending on the target product profile, different cassava breeding programs may have different cut-offs, mindful of the efficiency of traditional processing methods available and of local and international regulations on acceptable maximum residual levels of cyanide in cassava food and processed products.

3.2. *Effect of the environment on fresh root HCN content*

Fresh cassava root HCN content is known to vary even within roots of the same variety (10,49). These variations can in part be accounted for by differences in genotypes, climatic factors, and soils (12,18,50,51). The highest mean HCN by location was recorded in Arua (7.7) and Serere (7.58) (Table 2). These locations also had the highest temperatures recorded during the crop growing period (23.9 K and 22.99 K) respectively and the least precipitation and relative humidity after 18 hours (Figure 5B). This is consistent with observations by (31,52) who reported that cyanide in cassava roots tended to increase in periods of drought and/or dry weather.

The higher cassava root HCN levels recorded in Arua and Serere could be explained by the effect of temperature on the crop's physiology. Cyanide in cassava is produced from the breakdown of linamarine and lotaustraline, a process catalyzed by *hydroxynitrile lyase* (HNL) enzyme, which is in the leaves and stems but not in the roots (53). Increase in temperature increases rate of transpiration and sap flow in the plant, but could also increase the rate of HNL enzyme activity, thus elevating cassava root cyanide levels (53). Furthermore, temperature might influence the rate of cyanide

detoxification in cassava roots via the β – *cyanoalane* pathway, which converts cyanide to the amino acid asparagine. This reaction is catalyzed by β – *cyanoalane* synthase (CAS) enzyme, which is sensitive to heat denaturation, therefore its activity could be suppressed at high temperature (53).

Given the projected 1.2°C – 4.4°C increase in temperature on the African continent between the year 2030 and 2050 (54), it is essential that cassava varieties that are high yielding, low in HCN and stable across environments be developed. However, variations in temperature and relative humidity alone may not fully explain the effect of the environment on HCN phenotypic expression. In Tanzania, farmers claim that low fertility soils, red clayey and sandy soils have a tendency to produce bitter cassava roots (30). The effect of climatic factors and soil physical (texture) and chemical composition (pH and mineral composition) ought to be studied simultaneously across seasons and locations for more compelling observations.

BLUPS offer excellent predictive ability to guide selection decisions (55–57). Five of the bitter landraces featured among the top 10 high HCN clones, while one (Edwarat) was surprisingly ranked among the top 10 low HCN clones by two of the methods (Table 3). This clone is a landrace in a semi-arid environment in Eastern Uganda where it is said to be always bitter; thus, its name “Edwarat” which means “the bitter one.” This underscores the impact of environmental factors on fresh root HCN content. Cassava genotypes that are high yielding and low on HCN across several environments are highly desirable (50).

3.3. Accuracy of HCN phenotyping methods

The 1 to 9 scale (CV = 15.6%, $H^2 = 0.26$) is a more accurate and reproducible method compared to either the spectrophotometric method (CV = 24.6%, $H^2 = 0.14$) or the 0 to 800 scale (CV = 38.5%, $H^2 = 0.32$). Thus, whereas the spectrophotometric method is more accurate than the 0 to 800 scale, it is less reproducible. The 0 to 800 scale seems to be accurate for low HCN scores up to 300 ppm but is inaccurate for higher scores (Figure 3A). The spectrophotometric method on the other hand seems less accurate for the low values (less than 400 ppm) but accurate for the high values (Figure 3B). The large residuals the 0 to 800 scale could be due to the difficulty in distinguishing picrate papers colors and assigning the correct corresponding scores, an inherent problem associated with color scales (58). Furthermore, residuals observed when using the spectrophotometer could possibly be exacerbated by human error in time intervals colored picrate papers are eluted in water, especially when large sample numbers must be read on the spectrophotometer in a single batch. For the 1 to 9 scale, residuals were equally spread out across the range of predicted values, suggesting a constant variance of residuals across the scale (Figure 3C). The high coefficients of determination between the methods ($R^2 = 0.75 - 0.92$) suggest that one can predict phenotypes of one method, using another with a high degree of confidence since the methods yield comparable results.

For selection purposes, the 1 to 9 scale and spectrophotometric method outperformed the 0 to 800 scale as between them, they consistently selected 9 out of the top 10 low HCN clones. Given the extra requirements in sample preparation, machinery, and power, which make the spectrophotometric protocol an error prone, expensive, and lower throughput alternative, the more accurate 1 to 9 scale should be the method of choice for routine phenotyping in cassava breeding programs. At late-stage breeding trials when the population size is small and exact fresh HCN content is required to support variety release applications, the spectrophotometric method can be deployed. However, given the inherent difficulty in fresh cassava root HCN phenotyping, there is need to develop reproducible, accurate, and high throughput phenotyping protocols for HCN.

Near infrared spectroscopy (NIRS) is one such technologies that has been adopted by major cassava breeding programmes in Uganda and Nigeria for cassava root quality phenotyping (59-61) and has recently been shown to separate low and high HCN clones with high accuracy (62). Thus, cassava breeders targeting low HCN cassava varieties may find the concept of phenomic selection as recently described by (63) more appealing compared to genome based tools like marker assisted selection (MAS) given the phenotypic plasticity of the trait. Acyanogenic cassava can be produced via genetic transformation (64). However, in many African jurisdictions as is the case in Uganda, transgenics still face stiff resistance from anti-transgenics lobbyist groups and would be break

through research (65,66) is largely restricted to laboratory and confined field trials. Thus, conventional breeding for low cyanogenic cassava remains the most potent avenue to safeguard the public from potential harmful effects of dietary HCN consumption.

4. Materials and Methods

4.1. Description of the study area

Variability for HCN was studied across four locations; The National Crops Resources Research Institute (NaCRRI), central Uganda, the National Semi-arid Resources Research Institute (NaSARRI), Serere in eastern Uganda, Abi Zonal Agricultural Research Institute (Abi ZARDI), Arua northern Uganda and Tororo in eastern Uganda. NaCRRI, is located at longitude 32.62717 and latitude 0.521526 and 1134 meters above sea level, characterized by a bi-modal rainfall pattern with an average annual rainfall of 1325 mm and annual average temperature of 21°C. The soils are sandy loams. NaSARRI is located at longitude 33.54901 and latitude 1.499403 with an elevation of 1104 meters above sea level. The area is semi-arid with average annual temperatures of 26.5°C. The soils are light sandy loams. Abi ZARDI is located at longitude 30.95 and latitude 3.08333 at an altitude of 1109 meters above sea level, experiences a bi-modal rainfall pattern with average annual rainfall of 1250 mm and average annual temperature of 24°C. The soils are sandy loam. Finally, Tororo is located at longitude 34.0223 and latitude 0.769 with an elevation of 1278 meters above sea level. The mean annual temperature of 21°C. The area is characterized by light sandy soils and a bi-modal rainfall pattern.

4.2. Description of study materials and field trial establishment

A set of 64 clones was used in this study. These consisted of eight bitter landraces collected from farmers' fields across the country and 56 advanced breeding clones. These clones were selections from cycle two of genomic selection population (67). At each location, clones were established in the field in a randomized complete block design with two replicates. Each plot consisted of five rows of plants with a spacing of 1 x 1 meter, with 2-meter alleys between plots. Plots were kept weed free by hand weeding. No fertilizers and/or supplemental irrigation were applied. Further details of the trials can be found in the online open access data repository; Cassavabase at www.cassavabase.org.

4.3. Sample selection, preparation, and data collection

4.3.1. Sample selection

At harvest (12 months after planting), three middle rows of each plot were uprooted and all harvested roots pooled together. Three uniformly sized roots were then selected and taken to the laboratory for analysis. At the laboratory, roots were washed under running water to remove soil and debris and then dried with a towel.

4.3.2. Sample preparation

For each root, the distal and proximal portions were sliced off with a kitchen knife and discarded, leaving a 10-centimeter middle portion. This portion was peeled with a kitchen knife and the parenchyma grated with a kitchen grater. The grated slices per root were wrapped together and a portion immediately weighed off into a falcon tube containing 5 ml of pH 8 phosphate buffer. A fresh picrate paper strip (1 cm wide, 3 cm long) prepared by dipping Whatman 1 filter paper (Whatman International Limited) in a solution of picric acid (0.5% w/v) in 2.5% (w/v) sodium carbonate and allowing the paper to dry at room temperature was immediately suspended above the falcon tube and the tube tightly stoppered. The tubes were kept in the dark for 16 hours at room temperature.

4.4. Data collection

4.4.1. Measurement of HCN

Three assessments for HCN were undertaken. Firstly, scores were taken of the final color of the picrate paper strip on a 1 to 9 scale with 1 and 9 representing extremes of low and high HCN content respectively (34). Secondly, scores of the picrate paper were also taken using the 0 – 800 parts per million (ppm) scale with 0 ppm and 800 ppm representing extremes of low and high HCN content respectively (8). Thirdly, the scored picrate paper was eluted in 3 milliliters of distilled water for 30 minutes and the absorbance of the resultant-colored solution read on a spectrophotometer (Jenway 6305 UV/Visible Spectrophotometer) at 510 nm against a similarly prepared blank but whose picrate paper was not exposed to the sample. To get the final HCN value in parts per million, the absorbance value of the sample was multiplied by 396 (32).

4.4.2. Weather data

Weather data were obtained for the study sites from the AgERA5 database (68) using the R package *ag5Tools* (69). The AgERA5 dataset provides daily surface meteorological data from 1979 at a resolution of 0.1° (70). Thus, daily weather data were extracted for the period September 2020 to September 2021 for the parameters; 24-hour temperature (K), relative humidity after 12hours (%), relative humidity after 18 hours (%) and precipitation (mm/day).

4.5. Data analysis

For each of the three HCN assessment methods, an independent data set was assembled. Data analysis was done in R statistical package. Analysis of variance was done to test for significance of observed differences in means of clones, locations and for the interaction of genotypes with location (GEI) at 0.05 alpha level. Means of clones and locations were compared using Tukey's honestly significant difference test using the *HSD.test* function of the *Agricolae* package at 0.05 alpha level. Broad sense heritability was computed based on single site analysis and on a combination of data sets from all the sites. The *lmer* function of R statistical package was used to fit the mixed linear model;

$$y = X\beta + Zu + \varepsilon$$

where y is the response vector of HCN for a given location, β is the vector of fixed effects and u is the vector of random genetic effects with design matrix Z (relating trait values to genotype, environment, and genotype by environment interaction) and ε is the error (residual). Variance components were then extracted and used for the estimation of broad sense heritability (H^2). Within locations (single trial), broad sense heritability was estimated as;

$$H^2 = \frac{V_g}{V_g + V_e}$$

where; H^2 is the broad sense heritability, V_g is the genotypic variance, V_e is the error (residual) variance. Across locations (trials), broad sense heritability was estimated as;

$$H^2 = \frac{V_g}{V_g + V_{ge} + V_e}$$

where; V_g is the genotypic variance, V_e is the error (residual) variance, V_{ge} is variance due genotype by environment interaction

Ranking of clones according to HCN content was done using best unbiased linear predictors (BLUPS), which were computed using the *ranef* function in R. Variance components, heritability and BLUPS were computed separately for each HCN phenotyping method. The Kruskal-Wallis test was done to test the hypothesis that the test environments did not experience different weather conditions and results plotted using the *ggpubr* package. Correlation analysis was performed to investigate the relationship between the HCN phenotyping methods using the package *ggpairs* in R.

5. Conclusions

From this study; four conclusions are apparent; 1) there is appreciable genetic diversity for HCN in locally available Ugandan germplasm 2) heritability for HCN within the study germplasm is low to moderate 3) the 1 to 9 scale is more accurate and more reproducible than either the 0 to 800 ppm scale or spectrophotometric methods 4) average daily temperature and relative humidity during crop growth influence fresh cassava root HCN phenotypic expression. This information provides frameworks for undertaking sustainable cassava breeding mindful of acceptable thresholds for HCN in developed products and the phenotypic plasticity of the trait.

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Data Availability Statement: Data is available on the corresponding trial detail pages in an online open access repository; Cassavabase (www.cassavabase.org).

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