# Genetic diversity and population structure of *Peronosclerospora sorghi* isolates of Sorghum in Uganda

Kumi Frank<sup>1,2\*</sup>, Agbahoungba Symphorien <sup>1</sup>, Badji Arfang<sup>1</sup>, Mwila Natasha<sup>1</sup>, Ibanda Angele<sup>1</sup>, Anokye Michael<sup>2</sup>, Odong Thomas<sup>1</sup>, Wasswa Peter<sup>1</sup>, Ochwo-Ssemakula Mildred<sup>1</sup>, Tusiime Geoffrey<sup>1</sup>, Biruma Moses<sup>3</sup>, Kassim Sadik<sup>4</sup>, and Rubaihayo Patrick<sup>1</sup>

<sup>1</sup> Department of Agricultural Production, College of Agricultural and Environmental Sciences, Makerere University, P.O. Box 7062, Kampala, Uganda.

<sup>2</sup> Department of Crop Science, College of Agricultural and Natural Sciences,

University of Cape Coast, P.M.B Cape Coast, Ghana.

<sup>3</sup> National Semi-Arid Resources Research Institute (NaSARRI), Serere, Uganda.

<sup>4</sup>Abi-Zonal Agricultural Research and Development Institute, National Agricultural Research Organization, P. O. Box 219,

Arua, Uganda.

### Address of corresponding author: <a href="mailto:frankkumifk@gmail.com">frankkumifk@gmail.com</a>

**Abstract**— Sorghum is the third most important staple cereal crop in Uganda after maize and millet. Downy mildew disease is one of the most devastating fungal diseases which limits the production and productivity of the crop. The disease is caused by an obligate fungus, Peronosclerospora sorghi (Weston & Uppal) with varying symptoms. Information on the genetic diversity and population structure of P.sorghi in sorghum is imperative for the screening and selection for resistant genotypes and further monitoring possible mutant(s) of the pathogen. Isolates of P. sorghi infecting sorghum are difficult to discriminate when morphological descriptors are used. The use of molecular markers is efficient, and reliably precise for characterizing P. sorghi isolates. This study was undertaken to assess the level of genetic diversity and population structure that exist in P. sorghi isolates in Uganda. A total of 195 P. sorghi isolates, sampled from 13 different geographic populations from 10 different regions (agro-ecological zones) was used. Eleven (11) molecular markers, comprising of four Random amplified microsatellite (RAM) and seven (7) Inter-Simple Sequence Repeat (ISSR) markers were used in this study. The analysis of molecular variation (AMOVA) based on 11 microsatellite markers showed significant (P < 0.001) intra-population (88.9 %, PhiPT = 0.111) and inter-population (8.4 %, PhiPR = 0.083) genetic variation, while the genetic variation among regions (2.7 %, PhiRT = 0.022) was not significant. The highest genetic similarity value (0.987 = 98.7 %) was recorded between Pader and Lira populations and the

lowest genetic similarity (0.913 = 91.3 %) was observed between Namutumba and Arua populations. The mean Nei's genetic diversity index (H) and Shannon Information Index (I) were 0.308 and 0.471 respectively. Seven distinct cluster groups were formed from the 195 P. sorghi isolates based on their genetic similarity. Mantel test revealed no association between genetic differentiation and geographical distance ( $R^2 = 0.0026$ , p = 0.02) within the 13 geographic populations.

Keywords— AMOVA, Genetic diversity Index, ISSR, Mantel test, RAM, Shannon Information Index.

### I. INTRODUCTION

Downy mildew disease of sorghum, caused by an obligate soil-borne fungus Peronosclerospora sorghi (P. sorghi) [Weston and Uppal (Shaw)] (Frederiksen, 1980) poses a serious biotic stress to sorghum production and productivity worldwide (Williams, 1984). The prevalence and distribution of the disease is favoured by factors such as high relative humidity (Wang et al., 2000), low temperature (Bock et al., 1998) and rainfall which favours conidia production and subsequent development of the disease. When the disease is not managed at the early stages of infection, losses can reach 100 %. Infected plants show localized and systematic symptoms varying from lesions on leaf lamina to chlorosis (Jeger et al., 1998). Previous reports suggest the prevalence and distribution of Sorghum downy mildew disease (SMD) in Uganda at varying incidence and severity levels (Bigirwa et al., 1998; Kumi et al., 2018), but no study has been

done to unravel the state of genetic diversity of *P. sorghi* isolates in Uganda.

Morphological and molecular characterizations are two methods commonly used to assess genetic variability of *P. sorghi* (Bock, 1995). Unlike molecular characterization, morphological characterization is exclusively dependant on morphological traits of *P. sorghi* which come with several shortcomings ranging from; influence by environmental factors, time consuming and lack of efficient resolution power required to discriminate between genetically related isolates. More so, the use of morphological descriptors to assess genetic variation among isolates of *P. sorghi* is reported to be limited (Bock, 1995).

The advent of molecular markers has made it easier to study genetic diversity and population structure of *P. sorghi* with much precision and better accuracy (Perumal et al., 2006). Random Amplified Polymorphic DNA (RAPD) primers, Simple Sequence Repeats (SSRs) and Amplified Fragment Length Polymorphism (AFLP) have been extensively used to explore the variability among P. sorghi isolates (Bock et al., 2000) and different pathotypes of *P. sorghi* in sorghum (Perumal et al., 2008). Compared to other parts of the world, pathogenic and molecular variability among the P. sorghi isolates of sorghum have been well documented (Bock et al. 2000; Mathiyazhagan et al., 2008). Outcome of such research confirmed the existence of genetic diversity in P. sorghi and led to identification of new pathotypes (Perumal et al., 2006).

Perumal *et al.* (2006) analyzed the genetic variability among the 14 isolates of *P. sorghi* including metaxyl resistant isolates and reported that approximately 25% of the bands were polymorphic across the isolates in the tested populations. Mathiyazhagan *et al.* (2006) also reported similar genetic variability between isolates from sorghum and corn in India, using restriction fragment length polymorphism analysis of the polymerase chain reaction (PCR). Sequence characterized amplified region

(SCAR) marker have also been used to assess genetic variability of *P. sorghi* (Ladhalakshmi *et al*, 2009).

Knowledge regarding the extent of genetic diversity and genetic relationships of *P. sorghi* in Uganda will be valuable for designing a comprehensive breeding strategy for identifying resistant sorghum genotypes that will be resilient to SDM disease. Thus, studying and quantifying genetic diversity of *P. sorghi* from different agro-ecologies will offer a valuable marker assisted selection and breeding strategy. The objective of this study was to characterize the isolates of *P. sorghi* in Uganda by assessing the population structure, genetic diversity and/or relatedness using ISSR and RAMS markers.

# II. METHODOLOGY

### **Collection of samples**

Hierarchical sampling technique was used to collect a total of 195 SDM samples from 13 districts (Table 1 and Fig 1) covering all the ten agro-ecologies in Uganda. These samples were taken from sorghum plants with leaves showing systemic and/or localized characteristic lesions of mildew with conspicuous conidia growth. Leaves samples were carefully covered with aluminum foil, labeled and stored in a cooler.

### DNA extraction and purification

DNA extraction was done following the protocol described by McDermott *et al.* (1994). Conidia found on leaves were collected using a camel hair brush and transferred to a 1.5-ml microcentrifuge tube containing 500  $\mu$ l of extraction buffer (50 mM Tris-HCl, pH 8.0; 0.7 M NaCl and 1% SDS) (1 × 108 spores/ml), vortexed for 30 s and incubated at 60 ° C for 1 h. After incubation, the mixture was centrifuged at 12,000 × g for 10 min, and the aqueous phase was collected and extracted twice with an equal volume of phenol: chloroform: isoamylalcohol (25:24:1). The aqueous phase was transferred to a 1.5-ml microcentrifuge tube and the DNA was precipitated by addition of an equal volume of cold isopropanol and incubation at  $-20^{\circ}$ C for 1 h.

Populatio n	Latitude	atitude Longitude		Ave. Temp (° C)	Annual Rainfall (mm)	Agro-ecological zone
Iganga	N00°37.402'	E033°29.657'	1022	22.3	1313	Lake Victoria Crescent
Namutum				23.4	1011	Lake Victoria Crescent
ba	N00°48.309'	E033°39.273'	1091			
Pallisa	N00°48.309'	E033°39.273'	1081	23.2	1353	Lake Kyoga Basin
Kumi	N01°16.556'	E033°52.238'	1125	23.2	1238	Lake Kyoga Basin
Serere	N01°31.167'	E033°33.089'	1104	23.8	1362	Eastern Highlands
Lira	N00°18.478'	E032°34.292'	1153	23.6	1219	Northern Grassland

Table.1: Geographical and climatic data of P. sorghi populations in Uganda.

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	-					
Pader	N02°48.549'	E033°06.440'	979	23.7	1239	Northern Grassland
Hoima	N01°26.270'	E031°21.541'	1113	22.6	1382	Western Mid-Altitude
Masindi	N01°40.888'	E031°43.540'	1194	22.9	1355	Lake Albert Crescent
Arua	N03°01.303'	E030°54.684'	1109	22.9	1404	West Nile
Nebbi	N02°32.943'	E031°06.064'	1061	23.2	1098	Northwestern Grassland
Kabarole			1106	19.3	1459	Western Mid-High
	N00°36.511'	E030°15.428'				Altitude
Kabale	S01°15.660'	E030°01.444'	1934	17.2	1018	Southwestern Highlands

Source: Field data, 2016.



Fig. 1: Sampling sites of Peronosclerospora sorghi for the study. The small circled marks in each population (District) are the precised sampling points.

The DNA was pelleted by centrifugation at  $12,000 \times g$  at 4°C for 10 min. The pellet was washed twice with cold 70% ethanol, air-dried and re-suspended in 50 µl of Tris– EDTA buffer (10 mM Tris– HCl and 1 mM EDTA, pH 8.0). The genomic DNA was checked by agarose gel electrophoresis and the concentrations of the DNA were determined using a NanoDrop spectrometer nm (Thermo Scientific, Waltham, MA) by measuring the absorbance at 260.

Amplification and electrophoresis

A total of twenty (20) primers, consisting of nine (9) Random Amplified Microsatellites (RAMS) and eleven (11) Inter Simple Single Repeats (ISSRs) (Table 2) were used to screen the 195 *P. sorghi* isolates, but eleven (11) primers yielded clear visible PCR products and those were used in this study.

Primer	Primer sequence (5'-3')	Length	AT (°C)
UBC809 <sup>a</sup>	A GA GA GA GA GA GA GG	17	50
UBC824 <sup>a</sup>	TCTCTCTCTCTCTCG	17	47
UBC825 <sup>a</sup>	ACACACACACACACACT	17	50
UBC826	ACACACACACACACACC	17	47
UBC836 <sup>a</sup>	AGAGAGAGAGAGAGAGAGAGYA	18	50
UBC841 <sup>a</sup>	GAGAGAGAGAGAGAGAGAGATC	18	47
UBC842 <sup>a</sup>	GA GA GA GA GA GA GA GA	16	45
UBC847	CACACACACACACAAC	18	47
UBC848	CACACACACACACACG	18	45
UBC849	CTCTCTCTCTCTCTCA	18	47
UBC880 <sup>a</sup>	GGA GA GGA GA GGA GA	15	47
RAMS1 <sup>a</sup>	CACACAACAACAACAACA	18	45
RAMS2 <sup>a</sup>	GACCACCACCACCACCA	17	55
RAMS3 <sup>a</sup>	ATCCGACGACGACGACGA	18	50
RAMS4	AGGTGTGTGTGTG	14	45
RAMS5	AACACACACACA	12	35
RAMS6 <sup>a</sup>	ACCAGAGAGAGAGAG	13	40
RAMS7	GCATATATATATGT	14	35
RAMS8	CTAGAGAGAGCTCTG	15	45
RAMS9	GATCGCGCGCGCGTC	15	55

Table.2: List of Primers used in this study

a = Primers which yielded visible polymorphic bands.

These primers have the ability to differentiate among isolates of *P. sorghi*. Each PCR reaction (20  $\mu$ l) contained 50 ng of genomic DNA as template, 5 mM each dNTPs, 10 pmol of primer and 0.5  $\mu$ l of Taq DNA polymerase. DNA Amplification was performed using the following thermal cycle parameters; 94 °C for 5 min, followed by 30 cycles of 94 °C for 1 min, 37 °C for 1 min, and 72°C for 1 min and a final extension at 72 °C for 10 min at the end of amplification. The PCR products were analyzed by electrophoresis on a 1.5 % (w/v) agarose gel containing 0.5  $\mu$ g/ml of ethidium bromide in Tris-acetate-EDTA (TAE) buffer (0.04 M Tris-acetate, 0.001 M EDTA, pH

8.0) and subsequently visualized under UV light. Selected ISSRs and RAMs amplifications were further repeated to ensure reproducibility. Polymorphic bands were subsequently screened with a 100 base pair (bp) marker. **Data collection and analyses** 

Data collected from the amplified multiloci bands which were clearly visible were scored manually for each isolate using a binary format according to presence (1) or absence (0) (Santacruz-Varela et al., 2013) from the image of ethidium bromide stained gels (Fig. 2) of the target isolates.



M 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24

Fig. 2 Agarose gel electrophoresis of PCR amplified products from genomic DNA of P. sorghi using UBC 880 primer (ISSR primer) (a) Lane 1 (M) in both figure are 100 bp DNA ladder; Lane 2 – 24 (P. sorghi Isolates, Lane 14 is an empty well) and RAMS 3 (RAMS primer) (b) Lane 2 – 16 (P. sorghi Isolates).

The binary data were subjected to analysis of molecular variance (AMOVA) (Reyes-Valdés *et al.*, 2013; Excoffier *et al.*, 1992) with 999 permutations using GenAlEx6.5 (Peakall & Smouse, 2006; 2012). The total genetic variation among 195 isolates was generated using phi-statistic through AMOVA. The genetic variations were partitioned into three; variation among regions (PhiRT), variation among population (PhiPR) and variation within population (PhiPT).

PhiPT coefficient values denote the proportion of estimate variance among population relative to the total variance and the pairwise between populations expressed as PhiPT =  $(V_{AP} + V_{AR})$  ( $V_{WP} + V_{AP}$ ), where  $V_{AP}$  is the estimate of variance among populations,  $V_{AR}$  is the estimate of variance between the geographical regions, and  $V_{WP}$  is the estimate of variance within the studied population. PhiPT was used to determine the genetic differentiation between the population, it is a measure which allows intra-individual variation suppression when comparing binary and codominant data (Teixeira *et al.*, 2014).

The genetic variability of the populations

(districts) was analyzed using the Hardy-Weinberg equilibrium (HWE) (Crow et al., 2008) assumption in GenAlEx 6.5 (Peakall & Smouse, 2006; 2012). Pairwise Nei's genetic distances (Nei, 1972) and Pairwise Nei's genetic identity between geographical populations of P. sorghi were obtained based on 999 permutations. The percentage of polymorphic loci (PPL), number of bands (No. Bands), number of different alleles (Na), number of effective alleles (Ne), Shannon Information Index (I), expected heterozygosity/ Genetic Diversity Index (H) and unbiased expected heterozygosity (uHe). The Shannon diversity index (I) is an index that is commonly used to characterize species diversity in a giving population. Shannon's index (H) accounts for both abundance and evenness of the species present and expressed as a proportion of species i relative to the total number of species  $(p_i)$ , and then multiplied by the natural logarithm of this proportion  $(lnp_i)$ . The resulting product is summed across species, and multiplied by -1.

 $H = \sum_{i=1}^{s} - (pi * lnpi)$  (Spellerberg & Fedor, 2003)

Data on latitudinal and longitudinal coordinate for each isolate was converted into decimal degrees to

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estimate geographical distance matrix. Similarly, genetic distance was obtained by transforming the binary binary data (0 for absence and 1 for presence of amplicon) from the ISSR and RAMS amplified product. Mantel Test with 999 permutation was computed in GenAlEx to examine whether populations (districts) that are geographically close are also genetically similar.

Mantel test in GenAlEx (Peakall & Smouse, 2006; 2012) was performed to examine whether genetic isolation was associated with the geographic distance among P. sorghi populations. The pairwise Nei's population genetic distances were calculated based on gene frequency differences between populations, and these distances were then compared to geographic distances between populations and a correlation was run for these two parameters. The genetic distance matrix generated from 195 isolates in GenAlEx was used to perform a hierarchical cluster analysis based on Ward's hierarchical agglomerative clustering method (Ward, 1963) using R statistical package (R Core Team, 2013). A cluster dendogram was generated for the isolates (Maechler et al., 2016). In other to determine the optimal number of clusters for the dendogram, the Bayesian Inference Criterion (BIC) (Akogul, 2017) for k means method was used. This method deploys expectationmaximization, initialized by hierarchical clustering for parameterized Gaussian mixture models (Reynolds, 1992).

Following the cluster analysis, Discriminant Analysis of Principal Components (DAPC) (Jombart *et al.*, 2010) was carried out in R statistical package to assess the relationships between the different clusters, using a method that focus on variability between-group, while neglecting variability within-group variation, which is precisely the rationale of Discriminant Analysis (DA) (Lachenbruch & Goldstein, 1979) to achieve the best discrimination of isolates into pre-defined groups. DAPC scatter-plot was generated which allows for a graphical assessment of the genetic structures between clusters.

### III. RESULTS

Genetic variation and relationship in *P. sorghi* at varied population levels

Statistical analyses which were performed on the total of 120 amplified polymorphic loci from PCR reaction for *P. sorghi* isolates using ISSR and RAMS revealed a wide genetic diversity and structure in the population. The overall genetic differentiation was examined for the 195 *P. sorghi* isolates from 13 geographic populations (Districts) covering 10 regions (AEZ) in Uganda (Table 3). The AMOVA results are presented below.

populations of Peronoscierospora sorghi								
Sources	of	ďſ	MS	Est.	% Total	Phi	Value	P value
variation				variance	variation	Statistic		
Among		9	63.519	1.102	2.7	PhiRT	0.022 <sup>NS</sup>	0.487
regions								
Among		3	44.100	3.368	8.4	PhiPR	0.083***	0.001
population								
Within		182	22.468	35.876	88.9	PhiPT	0.111***	0.001
population								
Total		194		40.346	100			

Table.3: Analysis of molecular variance (AMOVA) showing the partitioning of genetic variation within and among nonulations of Peroposelerospora sorphi

Df= degree of freedom, SS= sum of square, Phi statistic, P value is based 999 permutation. \*\*\* = Significant at P < 0.001

Partitioning of genetic differentiation at three levels (among population, within population and among regions) contributed varying degrees of genetic variation to the total variation observed. Genetic variation among and within population of the 195 isolates examined by AMOVA were significant (P <0.001), while genetic variation of isolates among regions where not significant. Variation within population accounted for 88.9 % of the total genetic variance observed in *P. sorghi* isolates, this means the highest genetic diversity of these isolates occurred within population level. Variation among population and regions contributed 8.4 % and 2.7 % genetic variation respectively to the total diversity of *P. sorghi.* Additionally, significant (P <0.001) Phi values for genetic diversity were recorded among population (PhiPR = 0.083) and within population (PhiPT = 0.111) while the Phi value recorded among regions (PhiRT = 0.022) was not significant.

The results from the analysis of genetic diversity among 195 isolates of *P. sorghi* from13 different populations are presented in Table 4.

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Table	Table.4: Estimated Heterozygosity, number of bands and percentage polymorphism loci by Population									
Population	NPL	<b>PPL (%)</b>	Na	Ne	Ι	Η	uHe			
Kabarole	107	89.17%	1.783 (0.057)	1.462	0.436	0.285	0.294			
Arua	117	96.67%	1.942 (0.030)	1.564	0.493	0.329	0.341			
Nebbi	105	87.50%	1.750 (0.061)	1.378	0.381	0.243	0.251			
Iganga	109	90.83%	1.817 (0.053)	1.444	0.437	0.282	0.292			
Namutumba	105	87.50%	1.750 (0.061)	1.559	0.488	0.329	0.340			
Pallisa	113	94.17%	1.883 (0.043)	1.433	0.436	0.279	0.289			
Kumi	116	96.67%	1.933 (0.033)	1.502	0.485	0.316	0.327			
Kabale	112	93.33%	1.867 (0.046)	1.470	0.464	0.300	0.311			
Hoima	112	93.33%	1.867 (0.046)	1.599	0.525	0.354	0.366			
Masindi	116	96.67%	1.933 (0.033)	1.481	0.476	0.308	0.319			
Lira	119	99.17%	1.983 (0.017)	1.490	0.483	0.312	0.323			
Pader	119	99.17%	1.983 (0.017)	1.518	0.503	0.328	0.339			
Serere	118	98.33%	1.967 (0.023)	1.543	0.515	0.339	0.350			
Mean	113	94.04%	1.881	1.496	0.471	0.308	0.319			
Mean SE		1.17%	0.012	0.007	0.005	0.003	0.004			

PPL= Percentage of Polymorphic loci, NPL= Number of polymorphic loci, Na = Number of different Alleles, Ne = Number of Effective Alleles, I= Shannon's Information Index, H = Expected Heterozygosity/ Genetic diversity index, uHe = Unbiased Expected Heterozygosity. Values in parenthesis are standard errors.

The mean number of different polymorphic loci, percentage of polymorphic loci, number of different Alleles, number of effective alleles, Shannon's Information Index, Nei's genetic diversity index/expected heterozygosity (measure the number of alleles and their abundance) and unbiased expected heterozygosity were 113, 94.04 %, 1.881, 1.496, 0.471, 0.308 and 0.319, respectively. The results from the analyses revealed high levels of genetic variations within population.

The percentage of polymorphic bands was high in all the 13 distinct populations, with values ranging from 87.50 % for both Nebbi and Namutumba to 99.17 % also for both Lira and Pader. The number of different alleles ranged from a minimum of 1.750 recorded for population in Nebbi to a maximum of 1.983 for both Lira and Pader. The number of effective alleles ranged from a minimum of 1.378 for Nebbi population to a maximum of 1.599 for Hoima. Shannon Index was highest for Serere population (0.515) and lowest at Nebbi (0.381). The results show a low genetic diversity index (He) ranging from 0.243 for Nebbi population and 0.345 for Hoima population with an overall mean of 0.308.

# Relationship for genetic distance and geographical distribution

The results of pairwise Nei's genetic distance between geographical population of P. sorghi (above diagonal) and pairwise Nei's genetic similarity between geographical populations of P. sorghi (below diagonal) are presented in Table 5. The general Nei's genetic distance (genetic difference) values recorded between the tested populations of P. sorghi were very low, ranging from a minimum of 0.014 to maximum of 0.091. The smallest genetic distance was observed between Lira and Pader population (0.14), both of which are in the same agro-ecological zone (Northern grassland), while the largest genetic distance was observed between Arua and Namutumba population in which fall within West Nile and Lake Victoria Crescent ecological zones respectively. The entire population of *P. sorghi* isolates recorded high genetic similarity values ranging from 0.913 (91.3 %) to 0.987 (98.7 %). The minimum genetic similarity value of 91.3 % was recorded between Namutumba and Arua populations while the highest genetic similarity value of 98.7 % was recorded between Pader and Lira population.

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 Table.5: Pairwise Nei's genetic distances between geographical population of P. sorghi (above diagonal). Pairwise Nei's genetic similarity between geographical population of P. sorghi (below diagonal)

Populati	Kabar	Aru	Neb	Igan	Namutum	Palli	Ku	Kaba	Hoi	Masin	Lir	Pad	Sere
on	ole	а	bi	ga	ba	sa	mi	le	ma	di	a	er	re
		0.07	0.04				0.03				0.03	0.03	0.03
Kabarole	-	4	5	0.035	0.065	0.045	9	0.030	0.033	0.032	6	8	9
			0.08				0.06				0.07	0.06	0.06
Arua	0.929	-	8	0.071	0.091	0.086	5	0.068	0.059	0.072	4	7	9
		0.91					0.03				0.03	0.04	0.03
Nebbi	0.956	6	-	0.036	0.062	0.027	3	0.034	0.064	0.033	5	0	6
		0.93	0.96				0.03				0.03	0.03	0.03
Iganga	0.965	1	5	-	0.057	0.040	0	0.031	0.048	0.030	1	0	1
Namutu		0.91	0.94				0.07				0.08	0.06	0.06
mba	0.937	3	0	0.945	-	0.077	2	0.072	0.074	0.072	2	9	9
		0.91	0.97				0.03				0.02	0.02	0.02
Pallisa	0.956	8	4	0.961	0.926	-	0	0.031	0.050	0.027	2	6	4
		0.93	0.96								0.02	0.02	0.01
Kumi	0.962	7	7	0.971	0.931	0.970	-	0.016	0.045	0.017	2	2	9
		0.93	0.96				0.98				0.02	0.01	0.02
Kabale	0.970	5	7	0.969	0.930	0.970	5	-	0.047	0.016	1	9	2
		0.94	0.93				0.95				0.04	0.04	0.03
Hoima	0.967	3	8	0.953	0.929	0.952	6	0.954	-	0.040	2	1	7
		0.93	0.96				0.98				0.01	0.01	0.01
Masindi	0.968	0	7	0.970	0.930	0.974	4	0.984	0.961	-	8	9	8
		0.92	0.96				0.97					0.01	0.01
Lira	0.964	9	6	0.970	0.921	0.978	8	0.979	0.959	0.982	-	4	5
		0.93	0.96				0.97				0.98		0.01
Pader	0.963	5	0	0.970	0.934	0.974	9	0.981	0.960	0.981	7	-	6
		0.93	0.96				0.98				0.98	0.98	
Serere	0.962	3	5	0.970	0.933	0.976	1	0.978	0.963	0.982	5	4	-

The results from Mantel test to examine whether genetic distance was associated with geographic distance among *P. sorghi* populations revealed no association (Fig 4.  $R^2 = 0.0026$ , p = 0.02).



Fig. 4: Mantel test between Nei's genetic distance and geographic distance for 13 populations of P. sorghi (r = 0.014.)

#### **Cluster analysis**

The 195 P. sorghi isolates from 13 different geographic populations clustered into seven distinct

cluster groups (Fig. 4) which indicates genetic diversity within the population. Each cluster was mutually exclusive of isolates identity but not in geographical population of the isolates. The largest cluster was cluster 2 with a total of 64 isolates from 10 different geographic populations namely; Kabarole, Nebbi, Iganga, Pallisa, Kumi, Kabale, Masindi, Lira, Pader and Serere. The least weighted cluster was cluster 3 with 10 isolates from two geographic populations namely, Kabarole and Hoima. The remaining clusters, cluster 1, 4, 5, 6, and 7 constituted a total of 15, 25, 28, 31 and 22 isolates respectively from varied geographic populations. Cluster 1 constituted isolates from Kabarole and Namutumba while cluster 4 constituted population of Arua, Iganga, Kumi, Masindi and Pader. Furthermore, clusters 5 and 6 shared nine (9) geographical populations namely; Arua, Pallisa, Kumi, Kabale, Hoima, Masindi, Lira Pader and Serere. But in addition to the aforementioned populations, cluster 6 also Iganga. Lastly, cluster 7 consisted of populations namely; Nebbi, Iganga and Pader.



Fig 4: Ward's cluster dendogram of 195 isolates of P. sorghi from 13 different populations in Uganda.

Following cluster analysis and mantel test, Discriminant analysis was carried using the detected number of clusters from the dendogram computed from the Bayesian Information Criterion (BIC, which employs k means) Model to determine whether or not population structure exist within the *P. sorghi* isolates. DAPC results (Fig. 5) showed genetic variability among the isolates but there was no clear separation pattern among the study population. The results from the DAPC analysis showed no clearly defined population structure and a total of seven discriminant eigenvalues for principal components (PCs) were retained. The proportion of variance conserved by the PCA principal components accounted for 36.9 % variance. Discriminant analysis eigenvalues for PC1, PC2 and PC3 were 40.436, 17.523 and 9.166 respectively.

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Fig.5: Discriminant analysis of principal components (DAPC) for 195 P. sorghi isolates from Uganda. Each circle represents a cluster and each dot represents an isolate.

# IV. DISCUSSION

Genetic variability is significant to plant breeders for purposes of screening, selection and developing improved crop varieties that confers resistance to biotic stresses and are highly adaptable/ resilient to varied abiotic conditions while offering high quality yield and products for industrial purposes. In this present study, two different primer sets. Random Amplified Microsatellites (RAMs) and Inter-Simple Sequence Repeats (ISSRs) were used to assess genetic diversity and population structure of 195 P. sorghi isolates from 13 different populations in Uganda. The combination of these two molecular markers was effective in evaluating the genetic diversity, population structure and estimating genetic variations.

The observed genetic variation in this study was partitioned within population rather than among population or among regions (Table 3). These results indicate that majority of the genetic differentiation (88.9 %) existed within the defined population of *P. sorghi* isolates and among populations (8.7 %). However, the insignificant genetic differentiation value of *P. sorghi* isolates among regions (2.7 %) in this study showed that genetic differentiation was independent of the geographic regions (agro-ecological zone) of the isolate. These result agreed with other findings on genetic diversity in *P*. *sorghi* reported from other parts of the world such as India (Mathiyazhagan *et al.* 2008; Ladhalakshmi *et al.*, 2009), Indonesia (Lukman *et al.*, 2013) and the United States (Perumal *et al.*, 2008) and Africa (Bock *et al.*, 2000).

Results from this study also showed high percentage of genetic polymorphism (94.04 %) (Table 4) which explained the high genetic differentiation observed within the *P. sorghi* population in the AMOVA results. Similar results were reported by Sireesha & Velazhahan (2015) and Perumal *et al.*, (2008) who reported high genetic polymorphism in *P. sorghi* isolates from sorghum. In addition, Sireesha & Velazhahan (2015) reported high percentage polymorphic values for *P. sorghi* in sorghum which confirmed the high genetic variations observed in this study. High polymorphic percentage values recorded in this study (Table 4) further confirmed similar findings by Perumal *et al.*, (2008) and (Mathiyazhagan *et al.*, 2008) who also reported high percentage polymorphism (33 %- 100%) in *P. sorghi* isolates.

Low genetic diversity index (0.304) (Table 4) and high genetic similarity (98.7 %) (Table 5) values recorded in this study was not surprising because *P*. *sorghi* is reported to exhibit high sexual recombination (Heffer-Link *et al.*, 2002) and therefore explained the high level of inbreeding. Another contributing factor

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could be exchange of *P. sorghi* infected sorghum seeds among farmers for cultivation and thereby perpetuating the spread and development of the pathogen (seed-borne). These results were similar to findings of Sireesha & Velazhahan (2015) and Mathiyazhagan *et al.*, (2008) who reported high genetic simmilarity values of 93 % and 90 % respectively within *P. sorghi* isolates. Their results further reported low genetic diversity.

Mantel test results between geographic distance and genetic differentiation among the 13 geographic populations of *P. sorghum* revealed no significant correlation (Fig. 3). This result showed that, the observed genetic variability of *P. sorghi* isolates (within population) from the 13 different populations was not structured in geographic space (there is no spatial structure). Cluster analysis result (Fig. 4) revealed that, *P. sorghi* isolates in Uganda could cluster into seven (7) genetically distinct groups according to genetic similarity/identity of the populations. These findings suggested that geographical origin of *P. sorghi* isolates has no influence on the clusters formation.

Discriminate analysis result (Fig. 5) showed no clear well-defined pattern of clusters of genetic structure among the study populations. DAPC analysis therefore confirmed the Mantel test results, that the observed genetic variability of *P. sorghi* isolates within the study populations was not spatially structured.

# V. CONCLUSION

The study revealed a high genetic variation of *P. sorghi* within populations of Sorghi in Uganda. A weak association of genetic isolation by geographic distance was established for *P. sorghi* populations, which suggested that the observed genetic differences of *P. sorghi* was unaffected by the geographical populations. Seven genetically distinct clusters groups were formed from *P. sorghi* isolates according to the genetic similarities.

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