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# **Evaluation of population genetics of Moringa oleifera Provenances from Coastal Kenya using Single Nucleotide Polymorphism (SNPs) markers**

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#### **ABSTRACT**

*Population genetics analysis is a prerequisite to understanding how and why genotypes and allele frequencies and change over time between and within populations. Consequently, it offers insight into the process of evolutionary change and makes it possible to map variants linked to traits that differ among populations. In the present study, Single Nucleotide Polymorphisms (SNPs) markers were utilized to study the genetic characterization of 17 provenances from the Coast region of Kenya. 164 genotypes of Moringa oleifera were selected from 17 populations and genome sequencing undertaken utilizing genotyping by sequencing (GBS). Identification of polymorphisms (SNP Calling) in the selected genotypes and population genetic studies were carried out. SNP calling was done by Illumina's SNP caller algorithm in the CASAVA software. 20,921 SNPs were called with an average call rate of 0.82. Average polymorphsm content (PIC) for the SNPs was 0.24 and reproducibility was 0.98. A phenetic tree was constructed using a neighbor-joining approach using DArT R. For the population genetics analysis, F statistic (Fst) utilising the functions StAMPP package in DArT R was performed whereby Gede and Samburu exhibited the least heterozygosity/correlation with a value of 0.0003 whereas Pwani University and Samburu had the highest correlation of genes at 0.37. Euclidean was used as a measure distance, and the average distance between the populations was 33.024. The molecular variance analysis (AMOVA) described a lower, 2.55%, variation within the population and 2.73% among the populations. The high similarity between the genotypes could be attributed to the Moringa plants in the various provenances having the same ancestry. This study may help identify links between gene allelic forms and phenotypes, allowing the alleles to be connected to desired characteristics such as rapid growth and high yield (functional analysis), because of the high frequency of SNPs and their role as a source of allele variations.*

**Keywords:** Genotypes, Polymorphisms, Population Genetics

#### **I. INTRODUCTION**

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Genotyping by Sequencing (GBS) is a novel application to discover and genotype SNPs in crop genomes and populations utilizing NGS protocols for crop improvement (J. He et al., 2014). Based on genotyping-by-sequencing, DArT recently created a new genotyping platform called DArTseq which sequences the genomic DNA sample's most informative representations (Hassani et al., 2020), (Nadeem et al., 2018). Sample analysis for this study involved a highthroughput genotyping method using DArTseqTM technology from Diversity Arrays Technology Pty Ltd (Canberra, Australia). For next-generation sequencing, HiSeq2500 and DArT-SeqTM technology are utilized. DArTSeq, also referred to as DArT-Seq markers, is used to detect both SNPs and presence-absence sequence variations (Raman et al., 2014). New methodologies for sequence-based genotyping have been devised to put the power of NGS to work for plant breeding and genomics. Through the recent advancement of genotyping-by-sequencing (GBS) techniques, thousands of loci can be produced by sequencing a specific subset of the entire genome (Elshire et al., 2011). GBS is a feasible technology that utilizes enzyme-based complexity reduction (targeting only a small section of the genome with restriction endonucleases) with DNA barcoded adapters to create multiplex libraries of samples appropriate for NGS sequencing. This approach is known as reduced restriction sequence (RRS) (Wang et al., 2020). This technique has been shown to function across a variety of species and can generate tens of thousands to hundreds of thousands of molecular markers (Elshire et al., 2011). Given that *Moringa oleifera* is a cross-pollinated tree, significant variance in quantitatively inherited traits has been documented, and high heterogeneity across multiple character forms has been observed, leading to vast genetic diversity in both natural and cultivated accessions. These findings should be taken advantage of for focused research aimed at improving the crop. Comprehensive understanding of the gene flow pattern and population genetic structure of *M. oleifera* via molecular genetic diversity and global collection population structure holds significant potential (Lakshmidevamma et al., 2021). It is essential and necessary to characterize population structure and genome diversity across various sources for breeding programs to effectively manage germplasm (Hassani et al., 2020). Polymorphism detection at the genome level exposes the evidence of variation and polymorphism, allowing



for easy detection of genetic diversity within a population (Wu et al., 2010), (S.-T. He et al., 2021). Comprehending the genetic diversity within advanced breeding lines could then be used to enhance varietal selection and cross breeding (Kumar et al., 2017). Molecular markers are potent biotechnological tools that are frequently employed to evaluate genetic variations among plant species with the aim of differentiating genetically unique genotypes and create nutrientrich, high-yielding cultivars. (Sharma et al., 2024). The laborious process of developing markers can now be avoided by using sequences to in the detection and scoring of single nucleotide polymorphisms (SNPs) thanks to recent advancements in high-throughput sequencing (Chung et al., 2017). SNPs are the most recent molecular markers and can thus be utilized to detect genetic variations in plants as they are more efficient and prominent (Morgil et al., 2020). The present study aimed to undertake the molecular characterization 164 *M. oleifera* genotypes using DArTseq genotyping platform and thus evaluate the genetic population structure of the 17 provenances. As far as we are aware, this is the very first report for utilizing of SilicoDArTs and SNPs molecular variability for *Moringa oleifera* from the Coastal region of Kenya. In the future, genome-wide association mapping projects can utilize the generated DArTseq markers in the identification of QTLs and genes for traits that are desirable. This collection of markers will be extremely valuable for breeding programs.

#### **II. MATERIALS & METHODS**

#### **2.1 Collection of Plant Materials**

The original introduction sites from Kenya's Coast and some areas of its Eastern region were used to select provenances (fig. 1). This study was carried out in South Eastern Kenya University (SEKU) located 15 km off Kwa Vonza Market, along the Kitui-Machakos main road, Yatta/Kwa Vonza location, Lower Yatta Sub-County, Kitui County. The site lies at latitudes and longitudes of 037.755460 E 01.313580S respectively. The site elevation is 1173m a.s.l. as evident in figure 4 (Festus et al., 2022).164 most productive plants were selected (9 to 10 plants per provenance as seen in table 1) using a randomized complete block design (RCBD) (Festus et al., 2022).



*Location of Seed Collection Areas (Provenances)* (Festus et al., 2022)





### **Table 1** *No of Trees sampled per provenance*

### **2.2 Sample preparation**

Approximately 10 young and fresh *M. oleifera* leaves from the 164 sampled trees were selected and placed in a microtube. The samples were lypholized overnight at Seqart Africa Laboratories situated at International Livestock and Research Institute (ILRI), Kenya. The dried *M. oleifera* samples were then grinded using a Geno Grinder at 1300 rpm for 3 minutes. The resultant samples were utilized in DNA isolation (Ondieki et al., 2023)**.**

# **2.3 Genotyping by Sequencing**

## **DNA extraction and purification**

Genomic DNA isolation was performed using NucleoMag® 96 Plant according to manufacturer's protocol. Quantification of the isolated DNA was done using 0.8% Agarose gel to visualize bands.

# **2.4 Library Preparation**

A genomic representation of the samples was produced by combining two restriction enzymes (Pst1-CTGCAG- , Mse I) with adapters connected by barcode to identify each sample to function within a single lane on the Illumina HiSeq2500 device. The ends of the DNA fragments of interest were modified by adding DNA barcoded adapters and ligase to facilitate ligation with primers for amplification. MSE 1 was used as a common adapter and PST 1 enzyme as a barcode adapter. Effective PCR amplification was achieved at the MSE I site. 1.2% Agarose gel electrophoresis was used for quality control of the GBS library. The amplified fragments were combined in equimolar proportions using Tecan, an equipment designed for combining big samples. After applying the pooled samples to the Illumina c-Bot bridging PCR, a fully automated clonal cluster generation process, the Illumina HiSeq2500 system was used for sequencing. Sequencing libraries were prepared in compliance with SEQART Africa optimized protocols. The sequence DNA libraries were sequenced using an Illumina Hiseq 2500 (Ondieki et al., 2023).

# **III. DATA ANALYSIS**

The analytical program developed and patented by DArT Pvt. Ltd., Australia, was used to generate two types of data, scores for ''presence/absence'' markers (dominant), called SilicoDArTs and SNP markers (Spinoso-Castillo et al., 2020). Several quality parameters for DArTseq and SNP markers including call rate, polymorphic information content (PIC), and reproducibility, were computed by the DArTsoft v.7.4.7 (Hassani et al., 2020). This information was provided by DArT Pty Ltd to produce DArTSNP (table 4) and SilicoDArT (table 5). High quality markers were identified for this particular investigation by filtering the FASTQ data based on quality parameters, (Ondieki et al., 2023). Data resulting from the 164 genotypes was analyzed using KDCompute, a platform provided by Seqart Africa for various analyses and Diversity Arrays Technology (DArT) R, an R stats package developed by DArT. Reproducible

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and well-marked bands were scored for the presence (1) and absence (0) of single nucleotide polymorphisms (SNPs) using Gel Analyzer 2010a to form a binary data matrix (Spinoso-Castillo et al., 2020). Polymorphism information content (PIC) was calculated using silico DArT. Analysis of molecular variance (AMOVA) was computed based on the genetic distance matrix. For the population genetics analysis, F statistic (Fst) utilising the functions StAMPP package in DArT R was performed. Euclidean was used as a measure of distance. A phenetic tree was constructed using a neighbor-joining approach using DArT R. Analysis of molecular variance (AMOVA) was utilized to carry out population genetic analysis using KDCompute Software (silicoDArT) and DArT R.

## **IV. RESULTS**

## **4.1 To Identify Polymophisms in the Selected Genotypes**

A total of 37,853 markers (SilicoDArTs; table 5 and DArTSNPs; table 4) were scored using DArTsoft v.7.4.7 as binary data (1 present, 0 absent, for a marker in the genomic DNA of each accession). Several quality parameters for DArTseq and SNP markers including call rate, polymorphic information content (PIC), and reproducibility, were computed by the DArTsoft v.7.4.7. There were16,932 SilicoDArTs and 20,921 DArTSNPs markers.

The heterozygosity of the samples (FreqHet) ranged from 0 to 0.698 and exhibited an average heterozygosity of 0.11. The proportion of samples, which score as homozygous for the SNP allele (FreqHomSnp), ranged from 0 to 0.99 with an average of 0.2 (table 4). The proportion of samples for which the genotype score is "1" in the SNP allele row (OneRatioSnp) ranged from 0.006 to 1 with an average of 0.3. The average count of the reference, which is the sum of the tag read counts for all samples divided by the number of samples with non-zero tag read counts for the reference allele (AvgCountRef) ranged between 2.57 to 261.87 whereas its average was 24.96.

The average count of the SNPs, which is the sum of the tag read count for all samples divided by the number of samples with non-zero tag read counts for the SNP allele (AvgCountSnp) ranged between 2.5 to 181.97 and the average was 16.77. The proportion of technical replicate assay pairs for which the marker score is consistent (RepAvg) was 1. The proportion of samples for which the genotype score is "1" (One Ratio) was 1. The polymorphism information content (PIC) for the SNP allele row (PICSnp) ranged from 0 to 0.5 and exhibited an average of 0.24 whereas the average of the PIC of the Reference and SNP allele rows exhibited a range of 0.006 to 0.5 with average of 0.2 (table 2).

### **Table 2**



*DArT SNP marker quality and diversity*

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# **Table 3**

*Shows the SNP position*





# *Extrapolation showing part of the DArT SNPs*





*SilicoDArT data*



# **Bar plot on the Number of Individuals per Population**

Diani-Ukunda, India, Kibwezi Town, Kilifi, Likoni, Maungu, Mbololo, Miasenyi, Mwakiki, Pwani University and Shika Adabu provenances had 10 genotypes each whereas Arabuko Sokoke, Gede, Mackinon, Samburu, Vipingo and Waa Kwale had provenances had 9 genotypes as shown in fig. 2.



Vol. 5 (Iss. 4) 2024, pp. 836-851 African Journal of Empirical Research https://ajernet.net ISSN 2709-2607

# **Figure 2**

*Bar plot showing populations/ provenances*

# **F Statistics**

For Fst, the functions StAMPP package were used as they enable parallel computing hence faster. Fst nboots  $= 1$ , percentage was 95 and nclusters  $= 1$ . The genetic distances were in centiMorgans (cM)





*Fst table showing the distances in centiMorgans*





*Population genetic distances using Euclidean as a distance measure*No. of populations =  $17$ Average no. of individuals per population =  $9.647059$ No. of  $10c1 = 20921$ Miniumum Distance: 19.32 Maximum Distance: 48.45 Average Distance: 33.024 Completed: gl.dist.pop Arabuko-Sokoke Diani-Ukunda Gede India Kibwezi-Town Kilifi Likoni Mackinon Maungu Mbololo Diani-Ukunda 31,22686 Gede 33.60015 28.17896 India 31.70141 30.38961 30.71674 Kibwezi-Town 30.58207 28.96980 33.57191 31.67714 Kilifi 27.95676 23.67949 29.62305 28.59167 24.74401 Likoni 38.27676 31.25721 32.96569 34.65025 37.02305 31.45459 40.78608 45.22790 45.02551 Mackinon 44.88385 40.99392 39.02065 47.81913 26.02887 31.98760 31.13192 Maungu 29.98550 26.46278 20.04084 32.81997 40.22005 ofofodM 33.50289 28.30668 30.45371 31.23737 32.84962 27.40933 19.32103 45.07161 29.52481 32.64182 34.71259 35.31060 36.91502 32.88203 30.36716 48.45087 34.47238 31.28294 Miasenyi 37.11695 29.93681 31.63940 33.98191 35.71957 31.66914 23.32104 47.81128 33.60243 24.16349 Mwakiki 37.76208 37.65138 Pwani-University 37.59373 42.18504 40.48078 36.77890 31.72450 46.01462 43.07642 34.80793 41.99274 Samburu 30.90221 26.11825 25.45931 28.17443 30.49295 23.96683 30.11192 42.13376 27.73024 27.67423 Shika-Adabu 31,72473 29.76132 32.88128 31.63799 24.06153 27.99242 35.38716 43.29907 30.11396 31.49249 Vipingo 32.16829 25.28038 29.84218 30.68687 29.10405 22.95677 25.90581 41.36523 25.36065 24.02726 Waa-Kwale 35.33271 30.69813 32.56732 34.02535 33.65759 29.34749 30.16808 44.18125 30.99582 28.37661 Miasenyi Mwakiki Pwani-University Samburu Shika-Adabu Vipingo Diani-Ukunda Gede India Kibwezi-Town Kilifi Likoni Mackinon Maungu Mbololo Miasenyi Mwakiki 30.39267 Pwani-University 45.22906 46.18995 Samburu 33.20795 30.52589 38.62910 Shika-Adabu 36.20826 33.64458 39.53715 30.79956 31.88543 27.56765 36.96960 26.59748 30.34971 Vipingo Waa-Kwale 34.77817 31.58159 41.52538 29.99404 32.83590 28.69758

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# **4.3 Population Euclidean Distances**

The genetic distance between the 17 populations of *M. oleifera* genotypes were calculated and computed using DArT R package, based on Euclidean as a measure of distance. It was evident that the Euclidean distances ranged from 19.32 and 48.45 with an average distance of 33.024. The least distance was that of Mbololo and Likoni while the highest distance was between Miasenyi and Mackinon.

## **4.4 Analysis of Molecular Variance (AMOVA)**

Analysis of molecular variance (AMOVA) was utilized to carry out population genetic analysis using KDCompute Software (silicoDArT) and DArT R. The standard deviation and variance for individual alleles were calculated. The sum of the tag read counts for all samples, divided by the number of samples with non-zero tag read counts (AvgReadDepth) had a range of 255.26 to 5.05 with an average of 39.8. The standard deviation of the number of tag reads for all samples with non-zero tag read counts (StDevReadDepth) ranged from 0 to 124.14 exhibiting an average of 18. 96. The average of the normalized non-zero tag read counts divided by the standard deviation of the normalized non-zero tag read counts which is the variance (Qpmr), exhibited a range of 1.25 and 2262.8 with an average Qpmr of 2.58. Reproducibility ranged from 0.95 to 1 and had an average reproducibility of 0.99. From these data, it is evident that the average of the sum of the tag read counts for all the SNPs called was 39.8, the standard deviation of the SNPs was 18.96 whereas the variance of the SNPs was 2.58. From DArT R, the table below summarizes the values of analysis of molecular variance.

## **4.5 Neighbor-Joining Tree**

It constructs a phenetic tree using a neighbor-joining approach or UPGMA using DArT R and thus enables the visualization of the populations' genetic similarity. Euclidean distances were utilized. A phylogenetic tree with 17 tips and 15 internal nodes was resultant. It is evident from the phenetic tree (figure 3) that there were 4 clusters of the *M. oleifera* provenances. The 1<sup>st</sup> cluster had 4 populations. Pwani University and Mackinon populations were sub grouped together, indicating their close relationship. Maungu and Kilifi were also sub clustered together. In the  $2<sup>nd</sup>$  cluster were Shika Adabu and Kibwezi Town whereas in cluster 3 only Arabuko Sokoke was evident. Lastly, cluster 4 had 10 populations clustered together. Diani-Ukunda, India, Samburu, Gede, Waa Kwale, Vipingo, Miasenyi, Mwakiki, Mbololo and Likoni provenances were all clustered here. This shows their close genetic similarity.









## **V. DISCUSSION**

#### **5.1 To Identify Polymorphisms in the Selected Varieties**

As evident in Table 2, the allele with allele ID 100081123|F|0-33:T>C-33:T>C whose TGCAGAACAGCTACTCTTCCCTTGATATCCATCTTGGGCAGTCATATTCTGTCTTGGTCACAGCTAACC had a mutation or base pair change at position 33. The nucleotide change was from T to C hence the SNP was at position 33.

Another allele with Allele ID 100081125|F|0-19:G>A-19:G>A with the sequence TGCAGATAATTCGTCGTGCGAATGAGAATGGGGAAGATCCATTACATCTAAGCAACAGATTTTGTGAAG had a single nucleotide polymorphism at position 19 from G to A.

The call rate for the 20,921 SNPs ranged from 0.396 to 1 with an average call rate of 0.82. The proportion of samples for which the genotype score is "1" was averagely 0.30 with a minimum score of 0.006 and a maximum score of 1. The SNP exhibited a reproducibility of 0.98. There was an evidently high reproducibility and consistent marker score.

PIC for the SNPs was averagely 0.24 with a maximum PIC being 0.5. PIC is divided into three categories with regard to the polymorphism value, high (PIC value higher than 0.5), medium (value between 0.25 and 0.5) and low (lower than 0.25) (Vaiman et al., 1994; Xie et al., 2010)**.** In a study conducted by (Rajalakshmi et al., 2019), (Genetic Diversity, Population Structure and Correlation Study in *M. oleifera* Lam. Using ISSR and SRAP) Markers, fifteen inter-simple sequence repeat (ISSR) primers produced the average PIC value of 0.53 while the genetic variability study using sequence-related amplified polymorphism (SRAP) primer combinations exhibited an average PIC of 0.44. (Muslihatin et al., 2022) used ISSR markers to study the genetic diversity of *M. oleifera* where the average PIC value in the results of their study was 0.46. (kumar Ganesan et al., 2014) carried out Genetic diversity and population structure study of drumstick by utilizing morphological and SSR markers whereby the PIC value for SSR primers exhibited an average of 0.15 whereas the gene diversity ranged exhibited an average of 0.18. this was evidently lower than the current study. For the characterization of *M. oleifera*, (Saini et al., 2013) utilized RAPD, ISSR, and cytochrome P450 makers and reported high PIC values of 0.72, 0.81, and 0.68 for RAPD, ISSR, and cytochrome P450, respectively. It is evident that the average PIC for the SNP markers was lower. (kumar Ganesan et al., 2014) postulated that the reason for lower PIC could be due to the fact that SSR markers generated on average 1.84 bands per primer followed by minimum number of unique bands generated in the genotypes studied in comparison with (Muluvi et al., 1999) who reported 59 bands per primer with AFLP marker. In our study, this could also be attributed to SNP markers being changes on single nucleotide bases. (Sousa et al., 2017) found a mean PIC value of 0.35 with 11,187 SNP markers while studying *C. Arabica* and concluded that the low PIC value evidences the narrow genetic base of *C. Arabica*. Since PIC measures a marker's capacity to detect polymorphism, it can be assumed that the SNPs' low PIC average was brought on by the marker's specificity and selectivity. The heterozygosity of the loci ranged from 0 to 0.698 and exhibited an average heterozygosity of 0.11. This heterozygosity average was similar to that of (kumar Ganesan et al., 2014) average heterozygosity across 20 loci was 0.10.

The sum of the tag read counts for all samples, divided by the number of samples with non-zero tag read counts (AvgReadDepth) had a range of 255.26 to 5.05 with an average of 39.8. The standard deviation of the number of tag reads for all samples with non-zero tag read counts (StDevReadDepth) ranged from 0 to 124.14 exhibiting an average of 18. 96. The average of the normalized non-zero tag read counts divided by the standard deviation of the normalized non-zero tag read counts which is the variance (Qpmr), exhibited a range of 1.25 and 2262.8 with an average Qpmr of 2.58. Reproducibility ranged from 0.95 to 1 and had an average reproducibility of 0.99.

*M. oleifera* being a cross-pollinated tree, high heterogeneity in many character forms was observed resulting in vast genetic diversity. *M. oleifera* seems to be a very polymorphic plant since there were several single nucleotide polymorphisms observed. With an average call rate of 95%, 116,190 SilicoDArT markers were called on the 105 Napier grass accessions. The greatest anticipated heterozygosity (He) and polymorphism information content (PIC) values for biallelic markers were both 0.5. The expected heterozygosity (He) varied from 0 to 0.5, and the PIC ranged from 0 to 0.38. He and PIC values were generally 0.24 and 0.19, respectively. With an average call rate of 87%, a total of 85,452 SNP markers were called on the accessions. According to (Muktar et al., 2019), the PIC values ranged from 0 to 0.38 with an average of 0.11 and the He values ranged from 0 to 0.5 with an average of 0.13. This were higher than observed in the current study. The average PIC of DArT SNP for safflower was 0.29 (Hassani et.al,2020). This value was almost similar to the current study, whoseaverage PIC was 0.24.



#### **5.2 Population Genetics Analysis/ Study**

In-depth knowledge and understanding of the gene flow pattern and population genetic structure in *M. oleifera* through molecular genetic diversity and population genetic structure of worldwide collections is of immense promise. Documentation of genetic diversity and conservation of germplasm is a necessity to develop elite varieties.

#### **5.2.1 Population Euclidean Distances**

The genetic distance between the 17 populations of *M. oleifera* genotypes were calculated and computed using DArT R package, based on Euclidean as a measure of distance (table 7). It was evident that the Euclidean distances ranged from 19.32 and 48.45 with an average distance of 33.024. The least distance was that of Mbololo and Likoni while the highest distance was between Miasenyi and Mackinon. The genetic distances exhibited between the populations was relatively low and could thus be concluded that there was least variability in the populations. This could be attributed to the spread of seeds as most of the provenances are geographically close and the source of planting material arising from a common ancestry. According to (Mgendi et al., 2010; Muluvi et al., 1999),the core collection of *M. oleifera* populations extant in Africa consists of a relatively small number of accessions imported from India, which has led to low levels of genetic diversity.

### **5.2 F Statistics**

F statistics is used to test variances of populations (Gruber et al., 2018). Samburu and Gede exhibited the least variance 0.0003 whereas Pwani University and Mwakiki showed the highest variance with 0.37. A moderate Fst value would indicate minimal significant genetic divergence between the populations of *M. oleifera*, according to (Rufai et al., 2013). This study exhibited low Fst values between the populations signaling very low genetic divergence (table 6).

# **5.3 Neighbor Joining Tree**

The 1<sup>st</sup> cluster had 4 populations (figure 2). Pwani University and Mackinon populations were sub grouped together, indicating their close relationship. Maungu and Kilifi were also sub clustered together. In the 2<sup>nd</sup> cluster was Shika Adabu and Kibwezi Town whereas in cluster 3 only Arabuko Sokoke was evident. Lastly, cluster 4 had 10 populations clustered together. Diani-Ukunda, India, Samburu, Gede, Waa Kwale, Vipingo, Miasenyi, Mwakiki, Mbololo and Likoni provenances were all clustered here. This shows their close genetic similarity (Fig. 2).

Mackinon and Pwani University cluster together yet they are geographically distant. This observation is also true for Maungu and Kilifi, Samburu and Gede ad also Mbololo and Likoni. Hence, the conclusion that no geographic isolation was evident and could be resultant from the accessions having a common ancestry. This could also be explained from the perspective that no evolution has taken place to guarantee change in genetic makeup of *M. oleifera* since it was introduced less than a century ago by the Indians.

In the groups analyzed, no obvious geographic isolation was observed. (Rufai et al., 2013) and (Mgendi et al., 2010) reported a similar finding based on RAPD markers. This further demonstrates that population genetic divergence cannot be explained solely by geographic distance (Verma et al., 2015). Similar findings in other plant species have been reported by numerous studies (Singhet al., 2014; (Bronzini de Caraffa et al., 2002), but some of these studies were able to distinguish genotypes depending on geographic origin (Desplanque et al., 1999); (Muluvi et al., 1999) (Li & Nelson, 2002).

### **5.4 Analysis of Molecular Variance (AMOVA)**

AMOVA (analysis of molecular variance) was carried out to detect population differentiation utilizing molecular markers. The standard deviation and variance for individual alleles were calculated. The sum of the tag read counts for all samples, divided by the number of samples with non-zero tag read counts (AvgReadDepth) had a range of 255.26 to 5.05 with an average of 39.8. The standard deviation of the number of tag reads for all samples with nonzero tag read counts (StDevReadDepth) ranged from 0 to 124.14 exhibiting an average of 18. 96. The average of the normalized non-zero tag read counts divided by the standard deviation of the normalized non-zero tag read counts which is the variance (Qpmr), exhibited a range of 1.25 and 2262.8 with an average Qpmr of 2.58. Reproducibility ranged from 0.95 to 1 and had an average reproducibility of 0.99. From these data, it is evident that the average of the sum of the tag read counts for all the SNPs called was 39.8, the standard deviation of the SNPs was 18.96 whereas the variance of the SNPs was 2.58.

The molecular variance analysis (AMOVA) described a 2.55%, variation within the population and 2.73% among the populations. In this study the higher variation was among population. When AMOVA was used to analyze the SSR data, it was discovered that there was 2% variation between regions, 3% variation within populations, and 95% of the variation within populations (Singh et al., 2014). The majority of variation within populations is anticipated to be



maintained by woody, perennial, out-crossed species (Hamrick et al., 1989). In contrast to our finding that the greatest variation was seen between populations, (Muluvi et al., 1999) studied four pairs of AFLP primer combinations and found significant variation between regions and populations. However, (kumar Ganesan et al., 2014; Rufai et al., 2013), reported 95% of the total genetic variation within the population. The high similarity between the genotypes could be attributed to the *M. oleifera* plants in the various provenances having the same ancestry. It could also be resultant from the samples having been introduced from different sources.

## **VI. CONCLUSION**

*M. oleifera* seems to be very polymorphic as several SNPs were observed. This research sought to evaluate the population genetics of 17 provenances of *Moringa oleifera* from Coastal Kenya through molecular markers analysis, utilizing SNPs and genotyping.

It was also evident that the genotypes in the populations exhibited less variance. Population study showed that, no geographical isolation exists between genotypes from the Coastal region of Kenya. The less variability among the populations indicates that the populations are not significantly different genetically.

This could imply that spread of planting materials has taken place in the form of cuttings, seed and/or high rates of gene flow between the adjacent populations. This could also be explained from the perspective that no evolution has taken place to guarantee change in genetic makeup of *M. oleifera* since it was introduced less than a century ago by the Indians.

This study contributes significantly to the body of knowledge on *M. oleifera* genetics, paving the way for future research endeavors and practical applications aimed at harnessing the full potential of this remarkable plant. The management, conservation, association mapping, and marker-assisted selection of the drumstick for future improvement will all benefit significantly from this study.

## *Recommendations*

Continued research into functional genomics of *M. oleifera* will deepen our understanding of its biological mechanisms and facilitate development of innovative biotechnological applications.

More studies could be carried out using different molecular markers to ascertain the genetic divergence of Moringa in Kenya

## **Compliance with ethical standards**

*Disclosure of conflict of interest*

The authors declare that there is no conflict of interest in relation to this manuscript.

#### *Funding*

This research was funded by National Research Fund (NRF), Kenya. The authors acknowledge SEQART Africa, based at International Livestock and Research Institute (ILRI), Kenya for providing the laboratory and equipment for conducting this study.

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