Genome-wide mapping uncovers significant quantitative trait loci associated with yam mosaic virus infection, yield and dry matter content in White Guinea yam (Dioscorea rotundata Poir.)

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Introduction: Yam is an important crop for food security in East and West Africa due to its high market value and customer demand. High tuber quality with yield and disease resistance are the main traits for acceptability of yam cultivars across the tropical zone. There has been limited progress in enhancing the production and quality traits of yams, despite the significant socio-economic significance of this crop.

Method: To expedite the development of high-quality yam cultivars in Uganda, traits association study was conducted to identify genomic regions associated with key traits such as disease resistance, high yields, and dry matter content. The association mapping was conducted with multi-random mixed linear model (mrMLM) to compute the associations using five genetic models.

Results: A total of 16 significant single nucleotide polymorphisms (SNPs) markers were identified to be associated with the traits studied. Gene identification analysis revealed the presence of key putative genes such as Vicilin-like seed storage protein At2g28490 (ARATH)and Growth-regulating factor 1 involved in a variety of functions ranging from storage and gene regulation for disease resistance.

Discussion: The results obtained from this work have significant implications for the in-depth analysis of the genetic structure underlying key traits in yam. Additionally, this study emphasizes the identification of SNP variants and genes that may be utilized for genomic-informed selection in order to enhance yield and disease resistance in yams.

1 Introduction

Yam belongs to the *Dioscorea sp*ecies and plays a crucial role in ensuring food security in East and West Africa (Kilimo Trust, 2013). It is well-suited for market-driven production intensification because of its high market value and strong consumer demand (Haneishi et al., 2013). *Dioscorea rotundata*, often known as white yam or Guinea yam is a vital tuber crop with significant economic and dietary importance in several regions worldwide. The crop often known as the "*old man*" crop is cultivated by older individuals in Uganda (Adjei et al., 2022a). It is sometimes confused with the *Colocasia species* due to their similar tuber-producing nature (Kagoda et al., 2005). Over the past two decades, there has been limited progress in enhancing the production and quality of yams compared to other roots and tuber crops. This is particularly evident in East Africa, despite the significant socio-economic importance of yams. Nevertheless, in Uganda, yam breeding endeavors have not adequately explored the genetic foundation of valuable characteristics such as tuber quality and yield. This has hindered the rapid development of improved cultivars (Adjei et al., 2022a).. The improvement of several characteristics of yam such as resistance to pests and diseases, tuber production and quality is a challenging task due to their quantitative inheritance (Mignouna et al., 2001; Nemorin et al., 2012). While conventional breeding methods (Mass selection, phenotypic classification and hybridization) have been successful in improving many crop traits, there are certain traits that are not easily manipulated using these techniques. These traits often involve complex genetic traits influenced by multiple genes and environmental interactions (Acquaah, 2012).

The behavior of yams specifically in relation to their dry matter content (DMC), yield (TWY), and susceptibility to yam mosaic virus (YMV) has captured the attention of scholars, farmers, and breeders alike. To achieve sustainable yam production, provide food security and effectively manage diseases, it is critical to have a thorough understanding of these traits. The agricultural relevance and socio-economic effect of *D. rotundata* are greatly influenced by its dry matter content, yield, and sensitivity to yam mosaic virus (Mignouna et al., 2001). Continued research and innovative approaches in yam production are crucial for fully utilizing the potential of this important commodity. This is especially important as global concerns in food security and sustainable agriculture persist and evolve over time (Adeniji et al., 2012).

Dry matter content is a determinant of yam quality representing the remaining portion of the yam tuber after the removal of water content. The trait has a direct influence on the taste, consistency and nutritional composition of the yam. Due to its indication of a greater abundance of essential nutrients such as carbohydrates, a larger dry matter content is sometimes favored (Gatarira et al., 2020). To improve tuber quality and increase consumer satisfaction, researchers have been investigating the traits that influence dry matter content in Dioscorea rotundata. These traits include environmental conditions, cultivation methods, and genetic factors (Adjei et al., 2022b).. Yield, a significant trait of *Dioscorea rotundata* which constitutes the quantity of tubers generated within a given area (Wu et al., 2015). Optimizing yam output is important for meeting the food needs of growing populations and ensuring the economic prosperity of farmers (Adeniji et al., 2012). Yam productivity is influenced by crucial elements such as soil quality, planting density, pest and disease control, and proper watering. Scientists and agricultural experts are still exploring efficient methods to enhance yam production while maintaining environmental sustainability.

The cultivation of *Dioscorea rotundata* is significantly threatened by Yam mosaic virus (Mignouna et al., 2001). It is a viral disease that results in the formation of unique mosaic patterns on the leaves of yam plants. This leads to a decrease in the plant's ability to carry out photosynthesis, resulting in stunted growth and reduced output of yam tubers (Sorho Fatogoma Brahima Kone YKA& GOEDJB and Ettien Djecthi Jean Baptiste SFBKYKA& GO, 2014). Effectively managing YMV requires implementing a range of preventative strategies. These include using virus-free planting materials, adhering to proper sanitation practices, and employing cultivars that are resistant to the virus (Sorho Fatogoma Brahima Kone YKA& GOEDJB and Ettien Djecthi Jean Baptiste SFBKYKA& GO, 2014). Gaining insight into the interplay between *Dioscorea rotundata* and YMV is essential for devising efficient approaches to alleviate the adverse effects of this disease on yam cultivation (Adjei et al., 2022b).

The use of advanced breeding methods, such as genomic selection into yam improvement initiatives is anticipated to address complex challenges by enhancing the pace and effectiveness of breeding (Sugihara et al., 2020). The development of next-generation sequencing (NGS) technology has allowed for the discovery and utilization of single nucleotide polymorphism

(SNP) markers that are associated with certain traits. These markers may be used to aid in breeding programs for important crops (Bhattacharjee et al., 2013). A genome sequence assembly for *Dioscorea rotundata* has been recently published (Siadjeu et al., 2020). This development enables the utilization of high-density markers for Genome-Wide Association Studies (GWAS) to identify genomic regions associated with specific traits (Tamiru et al., 2017). Additionally, it allows for the accurate placement of SNPs at these specific locations. GWAS has been extensively employed to identify the intricate genes responsible for economic traits in crops. These qualities encompass disease resistance and quality improvement which are crucial for efficient breeding and genetic analysis (Cuevas et al., 2018; Ahn et al., 2019).

SNP Markers have been utilized in yams for a range of research, encompassing genetic analysis of tuber yield and yam mosaic tolerance (Agre et al., 2021), analysis of sex determination and cross-compatibility (Mondo et al., 2021), determination of paternity in yams (Norman et al., 2020), development of molecular markers (Girma et al., 2019), examination of yam hybrid origin at the genome level (Sugihara et al., 2020), and exploration of yam diversity (Loko et al., 2016; Agre et al., 2019; Bhattacharjee et al., 2020; Sugihara et al., 2020; Amponsah Adjei et al., 2023). The identification of favorable SNP alleles through QTL discovery is relevant for enhancing key traits via marker-assisted selection (MAS). Incorporating genomic-assisted breeding tools into yam breeding programs is anticipated to expedite genetic improvements for prioritized traits. This objective can be realized by elucidating the genetic mechanisms governing these traits, thereby enabling the systematic application of MAS for forward breeding.

Thus, this study aimed at identifying genomic regions associated with yield, YMV and dry matter content in *Dioscorea rotundata* species. The yam breeding program will benefit from the identified genomic areas in the marker-assisted selection process for the important traits under investigation. Furthermore, this study will establish a basis for genetic improvement and variety breeding for yams in Uganda by offering information for the investigation of high yielding, disease resistance-related, and high dry matter genes.

2 Materials and methods

2.1 Genetic materials

A total of 207 *D. rotundata* genotypes were used in this study. These populations consisted of breeding lines used in routine activities in Ghana, Nigeria, and landraces from Uganda. The genetic materials were part of the collection held at the National Crop Resource Research Institute (NaCRRI), which is an entity of the National Agricultural Research Organization (NARO) in Uganda (Supplementary Table 1).

2.2 Trials establishment and leaf sampling

The study was carried out in NaCRRI, Namulonge, Uganda, over two consecutive cropping seasons in 2020 and 2021. The location is situated at a latitude of 0°5` N and a longitude of 32°61` E. It has an elevation of 1,120 meters above sea level (masl) and experiences an annual rainfall of 1,170 mm (Nsubuga et al., 2011). The experiment was conducted using an augmented design, where each block (8 blocks) had 26 genotypes, three local checks, and three plants per genotype. The Genotypes were planted on mounds with inter-row and intra-row spacings of 1.2 m x 1.2 m, respectively. For each genotype, three pre-sprouted sets weighing between 400 and 500 g on average were planted in each mound. The plots were tagged for data collection. Yam leaf samples were gathered using the designated plant sample collection kit (KBS-9370-001) procedure. Using the BioArk Leaf sampling technique. Leaf samples were obtained from labeled plants and placed into 96-well tube plates 16 weeks following planting, utilizing a leaf puncher. Subsequently, the leaves were subjected to oven-drying at a temperature of 80°C.

2.3 Trait measurements

Data were collected for disease-related traits (Yam mosaic virus severity), yield-related traits (Yam tuber yield expressed as kg/plot) and dry matter content (%) (Table 1). All measurements were taken based on the standard operating protocol for the yam varietal performance evaluation trial (Asfaw, 2016) and the trait ontology dictionary described in YamBase (https://yambase.org/(accessed on: 18/12/2020) (Table 1).

2.3.1 Dry matter content

Dry matter content was determined seven days after harvesting by oven-dry method using a single tuber. Each tuber was sliced into chips, the fresh weight (wet sample) was taken and then oven dried. The oven for the experiment was preheated to a temperature of 80 ° C for 2 hours. After, the envelopes containing the sliced tubers were placed in the preheated oven for drying at 80 ° C for 48 hours. The dried sliced tubers were then weighed to determine the dry weight used as the dry matter content and expressed as a percentage (Equation 1).

$$\text{Dry matter content} = \frac{\text{weight of dry sample (g)}}{\text{weight of wet sample (g)}} \times 100 \qquad (1)$$

TABLE 1 Trait descriptors used for the evaluation of yam genotypes.

Descriptor	Description	Period of collection
Yam virus disease	1 = No visible symptoms, 2 = Mosaic on most leaves, 3 = Mild symptoms, 4 = Severe Mosaic and 5 = Severe leaf distortion and stunting	Monthly (8 weeks after planting)
Tuber weight	Calculated per plot	Between 1 to 14 days after harvesting
Dry matter content	Calculated using the oven method	

2.3.2 Yam mosaic virus estimates and area under disease progress calculation

The virus severity scores (Table 1) were averaged from three plant stands per plot. These averages were then used to calculate the area under the disease progress curve (AUDPC) values (Equation 2), following the method described by Forbes et al (Forbes et al., 2014)

The virus symptoms description in leaf included; 1 = No visible symptoms (virus negative), 2 = Mosaic on most leaves (symptoms recovery with time), 3 = Mild symptoms on few leaves (No leaf distortion), 4 = Severe mosaic on most leaves (leaf distortion) and 5 = Severe mosaic (bleaching/severe leaf distortion and stunting).

$$AUDPC = \sum_{i-1}^{n-1} \left(\frac{y_i + y_{i+1}}{2}\right)(t_{i+1} - t_i) \qquad (2)$$

Where;

- y_i = disease severity at the ith observation
- t_i = time (days) at the ith observation
- n = total number of observations

2.4 Genotyping and quality assessment

The dried leaf samples of each yam genotype were sent to SEQART AFRICA at the International Livestock Research Institute (ILRI) in Nairobi for genotyping. The process of DNA extraction was performed utilizing the Nucleomag Plant DNA extraction kit, specifically the Mag-Bind® Plant DNA DS 96 Kit. The isolated genomic DNA had a concentration ranging from 50 to 100 ng/μl. The quality and amount of DNA were assessed using 0.8% agarose gel. The libraries were generated utilizing the DArTSeq complexity reduction methodology (Kilian et al., 2016) by the digestion of genomic DNA using PstI and MseI enzymes. Subsequently, the barcoded adapters and common adapters were joined together, and then the resulting fragments were amplified by PCR. The libraries were subsequently subjected to single-read sequencing runs, with each run generating sequences of seventy-seven bases.

The Hiseq2500 platform was utilized to do next-generation sequencing (Kilian et al., 2016). The SEQART AFRICA platform utilizes genotyping by sequencing (GBS) DArTseqTM technology, which allows for rapid, high-quality, and cost-efficient genome profiling of complex polyploid genomes. The scoring of DArTseq markers was accomplished using DArTsoft14, an internal pipeline for marker scoring developed by Kilian et al (Kilian et al., 2016). The SilicoDArT markers and SNP markers were assessed using a binary scoring system. A value of 1 was assigned if the restriction fragment containing the marker sequence was present in the genomic representation of the sample, and a value of 0 was assigned if it was absent.

The SilicoDArT markers and SNP markers were mapped to the *D. rotundata* reference genome (TDr96_F1_v2_Pseudo Chromosome.rev07) in order to determine their placements on the chromosomes. The process of ensuring data quality and removing unwanted data was carried out using TASSEL (v5.2.52) (Bradbury et al., 2007). SNP markers exhibiting more than 20% missing data, a minor allele frequency (MAF) below 0.05, and places on the genome that are unknown were excluded. The SNP data underwent further imputation using the LD k-nearest neighbor genotype imputation (LD KNNI) approach (Troyanskaya et al., 2001), as described by Bradbury et al (Bradbury et al., 2007). This approach was selected because of Its ability to impute missing values by finding the closest neighbors in multidimensional space ensuring robustness considering the datasets used. Moreover, KNNI does not rely on assumptions about the underlying linkage disequilibrium (LD) structure, making it a flexible choice when the LD patterns are complex. In the end, a grand total of 4,957 Single Nucleotide Polymorphisms (SNPs) were chosen for subsequent examination.

2.5 Statistical analysis

2.5.1 Phenotypic data analysis

The phenotypic data collected from the two cropping seasons were analyzed using the "augmentedRCBD" function from the R package "agricolae" for the DAU test function (De Mendiburu, 2015). In the model, checks were treated as fixed while block and treatments were considered random. The adjusted means for the various genotypes in the two cropping seasons were derived through mixed model analysis (Equation 3) and were subsequently utilized in GWAS analysis (R Core Team, 2022).

$$\mathbf{Y} = \boldsymbol{\beta}_o + \Sigma\, \boldsymbol{\beta_i} \mathbf{Xi} + \Sigma\, \boldsymbol{\gamma_j} \mathbf{Cj} + \boldsymbol{\varepsilon} \qquad (3)$$

Where;

- **Y:** The phenotypic trait value.
- **β₀:** The intercept of the model.
- **βᵢ:** The effect size (regression coefficient) of the **i**th SNP genotype (typically coded as 0, 1, or 2 for the number of minor alleles).
- **Xi:** The genotype score for the **i**th SNP.
- **γj:** The effect size of the **j**th covariate.
- **Cj:** The value of the **j**th covariate for each individual (polulation structure).
- **ε:** The error term, representing all other unmeasured factors influencing the trait.

Σ represents summation across all SNPs (i) and covariates (j) included in the model.

In this case, we have as well the kinship matrix which is implemented in the mrMLM

2.5.2 GWAS Analysis and gene identification

A mixed linear model implemented in the multi-random mixed linear model (mrMLM) was used to compute the associations using five genetic models (Zhang et al., 2017). These models included: multi-locus random-SNP-effect Mixed Linear Model (Wang et al., 2016), fast multi-locus random-SNP-effect EMMA (FASTmrEMMA) (Zhang et al., 2020), polygenic-background-control- based least angle regression plus empirical Bayes (pLARmEB) (Zhang et al., 2017), fast

mrMLM (FASTmrMLM) (Tamba and Zhang, 2018) and pKWmEB (Wen et al., 2018).

The observed logarithms (-$\log_{10}$) of the p-values were plotted against the expected p-values to assess the adequacy of the GWAS model to examine how effectively the models compensated for population structure. The Manhattan plot was created for visualizing GWAS on the entire genome and zoom mapping was performed on a particular chromosome after identifying a significant SNP marker.

For gene identification, we made use of the generic feature format (GFF3) for searching for genes in the nearest associated marker (Hunter et al., 2012). Public database Interpro, European Molecular Biology Laboratory-European Bioinformatics Institute (EMBL-EBI) was used to determine the functions of the genes associated with the different SNPs identified (Shin et al., 2006).

2.6 Results

2.6.1 Genotypic variability for studied traits

The variance estimate for the traits under study was significant. For season one, the DMC varied from 15.9% to 45.5% with a mean of 30.9%, and for season two, it varied from 17.1% to 38% with a mean of 28.9% (Table 2; Figures 1, 2). When compared to the mean

of 5.5 (kg/ha) for season one, the mean yield was highest in season two (13.1 kg/plot). Additionally, season one had the lowest YMV mean incidence (175), ranging from 121.9 to 393.3, whereas season two had a mean of 217 and a range of 102 to 332.4 for YMV incidence. Heritability was highest for YMV in season one followed by DMC for the same season. The lowest was identified for tuber yield for both seasons respectively (Table 2).
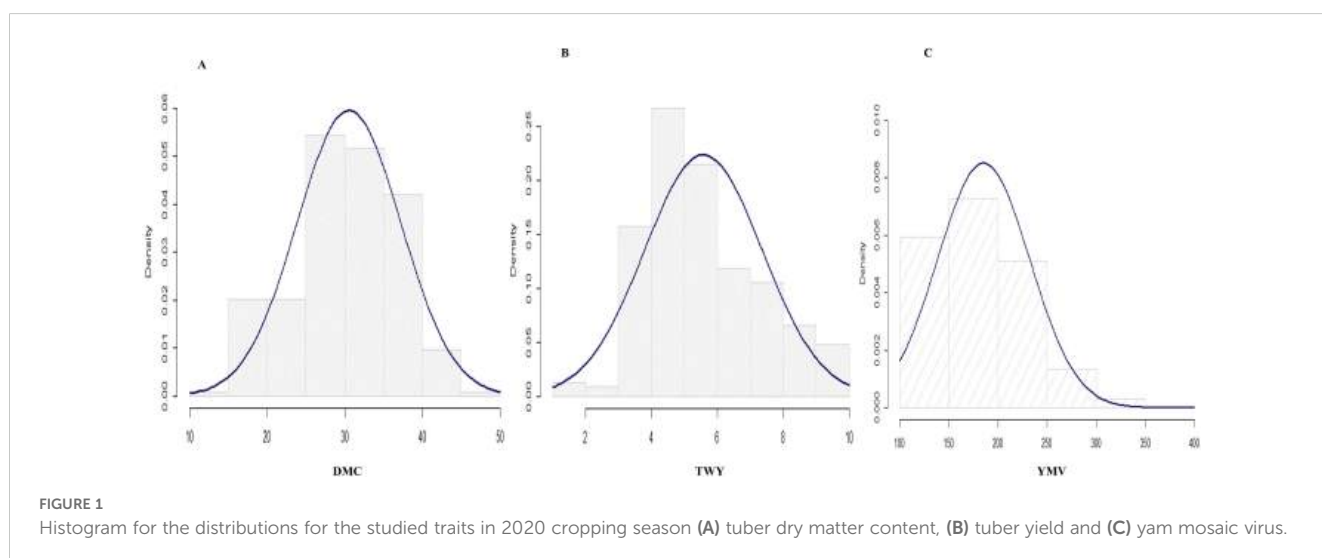
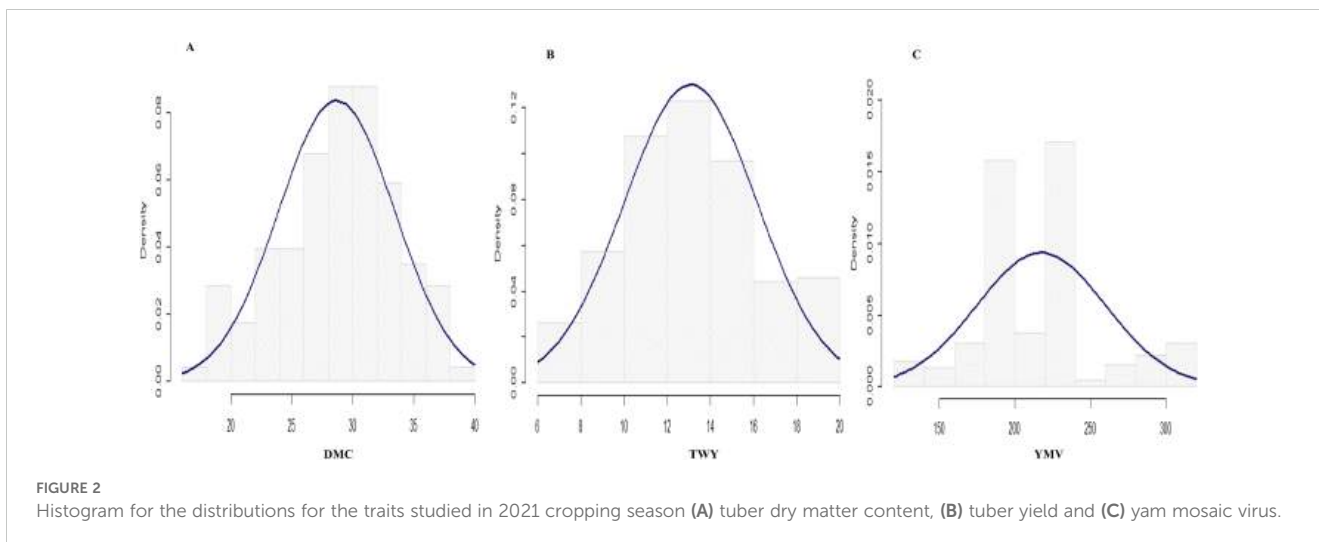### 2.6.2 Marker coverage, population structure and linkage disequilibrium

Table 3 provides the comprehensive count of single nucleotide polymorphisms (SNPs) detected on the chromosomes of *Dioscorea rotundata*, taking into consideration the absence of data, allele frequency, and heterozygosity. A total of 4,957 single nucleotide polymorphisms (SNPs) were retained and unequally distributed across the 20 chromosomes. Chromosome 5 had the highest number of SNPs, with 524 SNPs accounting for 10.6% of the total while chromosome 13 had the lowest SNPs, with 91 SNPs, making up 1.8% of the total. When comparing the observed and predicted heterozygosity, the observed heterozygosity varied between 0.097 and 0.153, with an average of 0.115. On the other hand, the expected heterozygosity ranged from 0.256 to 0.314, with an average of 0.280 (Table 3).

TABLE 2   Summary description for the traits studied.

Seasons	Traits	Minimum	Maximum	Mean ± SD[a]	Genotypic Variance	Phenotypic variance	Environmental variance	Heritability
One	DMC	15.9	45.6	30.9 ± 6.3	7.10	11.7	36.8	0.61
	TWY	1.5	9.9	5.5 ± 1.8	0.09	0.3	3.1	0.30
	YMV	121.9	393.3	175 ± 49.3	534.9	753.4	1747.8	0.71
Two	DMC	17.1	38.7	28.9 ± 4.8	2.20	4.5	18.3	0.48
	TWY	6.2	19	13.1 ± 3.8	0.30	1.2	9.4	0.25
	YMV	102	332.4	217 ± 45.6	189	258.5	1747.8	0.73

[a]SD, Standard deviation.

FIGURE 1
Histogram for the distributions for the studied traits in 2020 cropping season **(A)** tuber dry matter content, **(B)** tuber yield and **(C)** yam mosaic virus.

FIGURE 2
Histogram for the distributions for the traits studied in 2021 cropping season **(A)** tuber dry matter content, **(B)** tuber yield and **(C)** yam mosaic virus.

TABLE 3 Quality and summary statistics of DArTseq-SNPs on yam chromosomes.

Linkage group (Chromosome)	Filtered SNPs[a]	Observed heterozygosity	Expected heterozygosity	Polymorphic information content
1	166	0.101	0.213	0.239
2	152	0.111	0.229	0.256
3	213	0.097	0.229	0.254
4	367	0.098	0.216	0.254
5	524	0.080	0.235	0.264
6	188	0.105	0.223	0.243
7	251	0.110	0.232	0.254
8	320	0.125	0.229	0.268
9	205	0.142	0.252	0.275
10	166	0.113	0.216	0.238
11	210	0.110	0.234	0.256
12	202	0.153	0.253	0.288
13	91	0.112	0.231	0.253
14	305	0.140	0.257	0.276
15	249	0.117	0.234	0.257
16	204	0.110	0.248	0.272
17	253	0.132	0.238	0.261
18	244	0.123	0.239	0.251
19	485	0.115	0.246	0.265
20	162	0.112	0.220	0.250
Total/Average	4957	0.115	0.234	0.259

[a]SNPs, Single Nucleotide Polymorphisms.

A principal component analysis (PCA) was conducted utilizing the pairwise Euclidean genetic distance matrix of the genotypes to illustrate the genetic divergence in the yam genotypes. The PCA findings demonstrated that the first two axes accounted for 99.7% of the overall genetic variance (Figure 3A). The first axis (Dim 1) accounted for 94.9% of the genetic variance, while the second axis (Dim 2) accounted for 4.8% of the overall genetic variation. The genotypes, as determined by SNP markers and principal component analysis, exhibited limited diversity with their geographical origins. The graph displayed three prominent clusters, revealing that the genotypes obtained from Nigeria and Ghana formed a cohesive group, whereas the genotypes from Uganda were dispersed among the genotypes from Nigeria and Ghana in various quadrants (Figure 3A).

The LD plot indicated a significant association between linkage disequilibrium (R^2) and physical distance (bp) (r = −0.035), as well as between p-value and R^2 (r = −0.40), suggesting the presence of linkage decay. The decrease of linkage disequilibrium (LD) varied

among the chromosomes, with a range of 8,289 base pairs for chromosome 1 to 58,562 base pairs for chromosome 20 (Figure 3B).

2.7 Genome-wide association scans for studied traits

2.7.1 Dry matter content

Based on the methods used for the study, we identified two significant SNP markers associated with the dry matter content, with all the SNP markers located on Chromosome 18 (Table 4; Figures 4, 5). The SNP loci exhibited marker effect ranging between 1.74 and 3.49 and together accounted for an average of 25.9% of the overall phenotypic variance with an average minor allele frequency of 0.212 for the two seasons. Additional analysis of the two SNP loci linked to the tuber DMC on chromosome 18 revealed three linkage alleles (AA, AT and TT). The association between loci AT and AA for marker was observed significant at $p<0.45$ compared to linkage between the AA
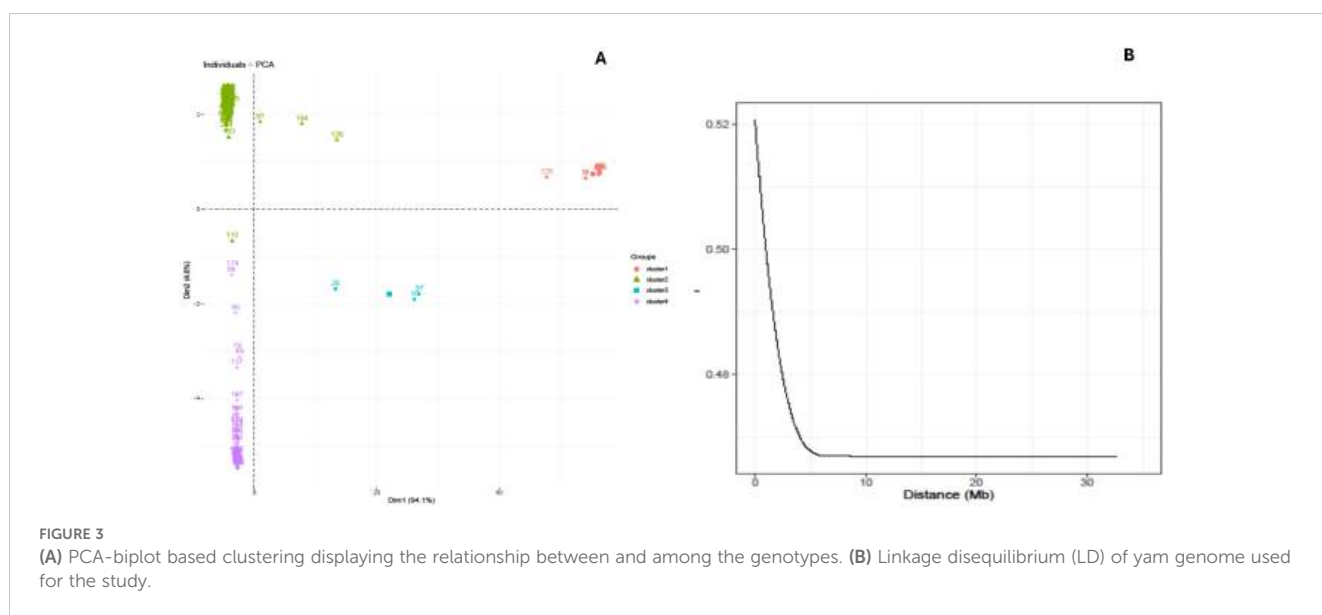

FIGURE 3
(A) PCA-biplot based clustering displaying the relationship between and among the genotypes. **(B)** Linkage disequilibrium (LD) of yam genome used for the study.

TABLE 4 Summary of significant single nucleotide polymorphism describing different genomic regions associated with studied traits for two cropping seasons in a panel of 207 Dioscorea rotundata genotypes.

Season	Trait	Method	Markers	Chr	Pos (Mp)	Effect	log10 (P)	r2 (%)	MAF	Var_error
One	DMC	pLARmEB	chr18_24123303	18	24.1	2.34	4.13	3.09	0.10	35.83
	TWY	FASTmrEMMA	chr4_21170357	4	21.2	-2.50	4.01	13.03	0.17	3.78
	YMV	FASTmrMLM	chr15_5690650	15	5.7	0.00	4.01	0.00	0.19	2177.76
Two	DMC	mrMLM	chr18_21626748	18	21.6	2.06	5.49	9.15	0.24	20.80
		FASTmrMLM	chr18_21626748	18	21.6	1.75	5.78	6.58	0.24	21.05
		FASTmrEMMA	chr18_21626748	18	21.6	3.50	4.09	7.15	0.24	21.05
		pLARmEB	chr18_21626748	18	21.6	1.75	6.10	6.54	0.24	21.05
		pKWmEB	chr18_21626748	18	21.6	1.75	4.91	9.15	0.24	21.05
	TWY	mrMLM	chr15_20074744	15	20	10.72	5.57	37.08	0.05	25.62

(Continued)

TABLE 4 Continued

Season	Trait	Method	Markers	Chr	Pos (Mp)	Effect	log10 (P)	r2 (%)	MAF	Var_error
		FASTmrEMMA	chr12_18361897	12	18.4	22.07	7.07	39.80	0.06	22.07
		FASTmrEMMA	chr14_7054136	14	7	17.16	7.50	25.89	0.07	23.04
		FASTmrEMMA	chr15_20539155	15	20.5	3.52	4.80	1.97	0.46	25.07
		pLARmEB	chr15_20539155	15	20.5	1.68	4.80	5.17	0.46	26.83
	YMV	mrMLM	chr9_96188	9	0.9	68.85	4.63	67.91	0.29	1793.14
		FASTmrMLM	chr19_19130862	19	19.1	24.69	5.56	2.22	0.08	1640.75
		FASTmrEMMA	chr15_262823	15	0.2	86.96	4.17	21.26	0.07	1730.48

and TT which was at $p < 0.054$. These alleles were observed in clones that possess the homozygous allele TT and AA exhibited high DMC compared to those with the heterozygous allele AT.

2.7.2 Total weight of yam

Six significant single nucleotide polymorphism (SNP) indicators related to yam tuber yield were discovered on four chromosomes (4, 12, 14 and 15). The following Markers "chr4_21170357", "chr12_18361897", "chr14_7054136" and "chr15_20539155"

located on chromosome 4, 12 14 and 15 showed a significant association using the FASTmrEMMA method (Table 4). The MAF ranged between 0.05 and 0.45, and they were detected with marker effect ranging between -2.50 and 22.06. These markers explained approximately 32.2% of the total phenotypic variation. For TWY, a total of 5 alleles (CC, CT, TT, AA and AT) were identified and alleles CC and TT are associated with high tuber yield (TWY) in the population. Conversely, the alleles AA, CT, and AT were found to be associated with low TWY.
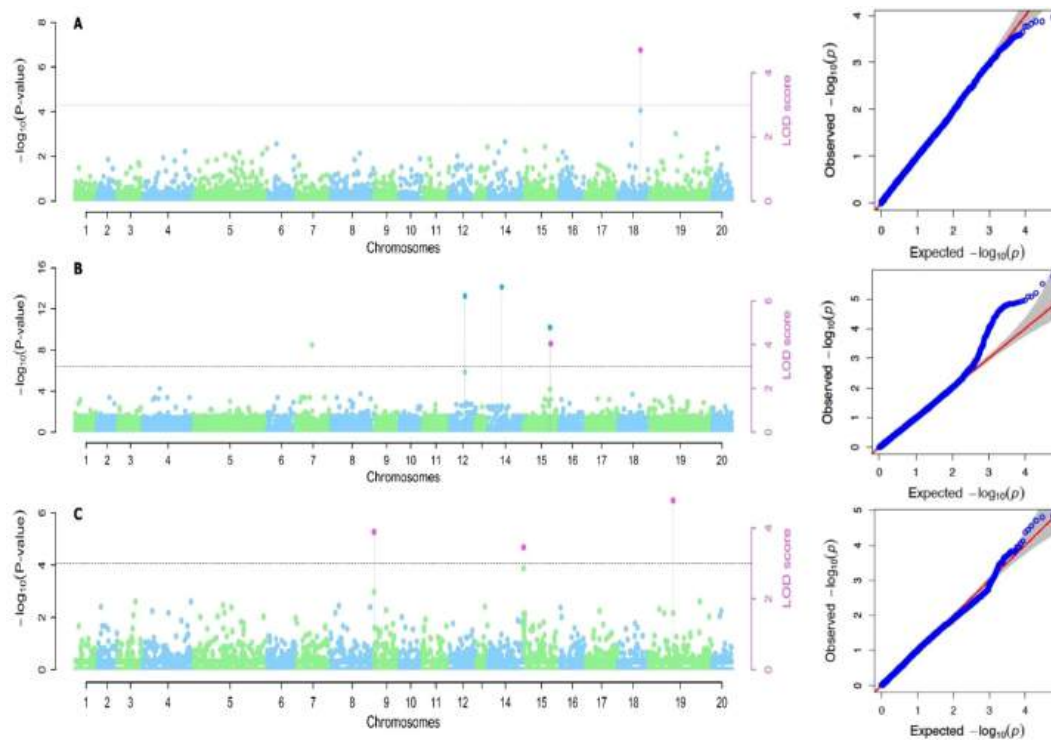

FIGURE 4
Manhattan plots for genome-wide diagnosis of association signals for studied traits of cropping season one [**(A)** Dry matter content, **(B)** Total weight of yam and **(C)** Yam mosaic virus]. [The y-axis represents the p-value of the marker-trait association on a -log₁₀ scale and the x-axis relates to the 20 yam chromosomes, green dots above the blue dotted line and red dot above red line indicate SNPs associated with QTL with influence on yam mosaic virus disease. Horizontal blue line is 5% Bonferroni threshold line].

FIGURE 5
Manhattan plots for genome-wide diagnosis of association signals for studied traits of cropping season two [**(A)** Dry matter content, **(B)** Total weight of yam and **(C)** Yam mosaic virus]. [The y-axis represents the p-value of the marker-trait association on a -log$_{10}$ scale and the x-axis relates to the 20 yam chromosomes, green dots above the blue dotted line and red dot above red line indicate SNPs associated with QTL with influence on yam mosaic virus disease. Horizontal blue line is 5% Bonferroni threshold line].

2.7.3 Yam mosaic virus

A total of four significant SNPs were found to be linked with YMV. Among these, two SNP loci were located on chromosome 15, and one was located on chromosome 19 at a physical position of 19,130,862 Bp. With FASTmrEMMA method, we identified three SNPs (chr15_5690650, chr15_262823 and chr19_19130862). Additionally, through the mrMLM method, we observed one SNP (chr9_96188) (Figures 4, 5). The SNP loci exhibited marker effect ranging from 24.68 to 86.96, and together accounted for an average of 21.12% (Table 4). The Quantile-Quantile (QQ) plot confirmed a decrease in the -log10 (p-value) toward the predicted level for YMV. In the case of YMV, the presence of the homozygous alleles CC and AA resulted in lower resistance compared to the heterozygous allele AT.

2.8 Gene identification

2.8.1 Dry matter content

Putative genes associated with dry matter content were mainly identified on chromosomes 18 (Table 5). Through the annotation process, two candidate genes linked to DMC were identified. The genes mentioned include Transcription initiation factor TFIID subunit 12b, and Growth-regulating factor 1. The gene Transcription initiation factor TFIID subunit 12b was found at position 24Mb on chromosome 18. The growth-regulating factor 1

gene was found at positions 21.6 Mbs on chromosome 18 for five methods used for the study (Table 5).

2.8.2 Total weight of yams

We identified putative genes associated with total tuber yield of which were identified on 6 chromosomes (Table 5). Among the 6 chromosomes, 5 potential candidate genes were found near the peak SNPs. These genes include Phoenix dactylifera coatomer subunit delta-2-like observed on chromosome 4 for positions 21.2 Mbs, COMPASS-like H3K4 histone methylase component WDR5AARATH (Chromosome 15; position 20.0 Mbs), Vicilin-like seed storage protein At2g28490ARATH (Chromosome 12; position 18.4 Mbs), Ananas comosus IAA-amino acid hydrolase ILR1-like 3 (Chromosome 14; position 7.1 Mbs) and Late exocytosis, associated with Golgi transport (Chromosome 15; position 20.5 Mbs), (Table 5).

2.8.3 Yam mosaic virus

However, five potential genes were identified for YMV, specifically the Glucan endo-1,3-beta-glucosidase 8 which is observed on chromosome 15 at 5.7 b. The second association was identified as chromosome 9 at 9Mb for nuclear pores complex protein NUP62 gene. The two genes (Cold-responsive protein kinase 1 and Probable LRR receptor-like serine/threonine-protein) were linked to chromosome 19 at 19.1Mb. The last gene

TABLE 5 Gene identification for the significant SNPs for studied traits.

Season	Traits	Method	Chr	Pos	GENE ID	Pfam	Function
One	DMC	pLARmEB	18	24,123,303	DRNTG_30791	PF03847	Transcription initiation factor TFIID subunit 12b
	TWY	FASTmrEMMA	4	21,170,357	DRNTG_22047	PF00928	Phoenix dactylifera coatomer subunit delta-2-like
	YMV	FASTmrMLM	15	5,690,650	DRNTG_12640	PF07983	Glucan endo-1,3-beta-glucosidase 8
Two	DMC	mrMLM	18	21,626,748	DRNTG_27518	PF08879	Growth-regulating factor 1
		FASTmrMLM	18	21,626,748			
		FASTmrEMMA	18	21,626,748			
		pLARmEB	18	21,626,748			
		pKWmEB	18	21,626,748			
	TWY	mrMLM	15	20,074,744	DRNTG_08898	PF00400	COMPASS-like H3K4 histone methylase component WDR5A(ARATH
		FASTmrEMMA	12	18,361,897	DRNTG_00658	PF00190	Vicilin-like seed storage protein At2g28490(ARATH)
		FASTmrEMMA	14	7,054,136	DRNTG_20660	PF07687	Ananas comosus IAA-amino acid hydrolase ILR1-like 3
		FASTmrEMMA	15	20,539,155	DRNTG_03009	PF13967	Late exocytosis, associated with Golgi transport
		pLARmEB	15	20,539,155			
	YMV	mrMLM	9	96,188	DRNTG_10092	PF05064	Nuclear Pore Complex Protein NUP62
		FASTmrMLM	19	19,130,862	DRNTG_03895	PF00069; PF07714	Cold-responsive protein kinase 1; Probable LRR receptor-like serine/threonine-protein
		FASTmrEMMA	15	262,823	DRNTG_23443	PF07714	G-type lectin S-receptor-like serine/threonine-protein

observed to be associated with YMV was G-type lectin S-receptor-like serine/threonine-protein for chromosome 15 at 0.2 Mb. The heatmaps display the linkage disequilibrium (LD) of each detected SNP locus. Regarding DMC, the LD analysis of the loci (traits under study) indicated that these markers exhibited a moderately to highly significant LD parameter ($R^2 > 0.6$), indicating a rather strong connection (Supplementary Figure 1). The LD analysis conducted on two loci, TWY and YMV, revealed that the markers on chromosome 5 exhibited very low LD characteristics ($R^2 < 0.7$) (Supplementary Figure 1).

3 Discussion

The study revealed an extensive genome-wide association analysis to pinpoint quantitative trait nucleotides associated with key agronomic traits in yam, such as Yam Mosaic Virus resistance, yield, and dry matter content. The analysis uncovered significant and informative variances among these traits. Employing multi-random mixed linear model approaches in mrMLM, five different gene action models were utilized. Across two seasons, a total of 16 SNP markers were identified. Similar chromosomal regions linked to yield and dry matter content were reported in previous studies (Agre et al., 2021).

This study identified four gene/protein families linked to yam mosaic virus disease: Glucan endo-1,3-beta-glucosidase 8 (Henrissat and Davies, 2000; Mouyna et al., 2000; Barral et al., 2004), Nuclear pore complex protein NUP62 (Bailer et al., 2001), Cold-responsive protein kinase 1; Probable LRR receptor-like (Manning et al., 2002; Li et al., 2004; Stout et al., 2004) and G-type lectin S-receptor-like serine/threonine-protein (Hanks and Quinn, 1991). These genes are located on chromosomes 9, 15 and 19, respectively.

According to Barral (Barral et al., 2004), the X8 domain contains at least 6 conserved cysteine residues that presumably form three disulphide bridges. The domain is found in an Olive pollen allergen as well as at the C-terminus of several families of

glycosyl hydrolases and observed to be involved in carbohydrate binding. It is characteristic of GPI-anchored domains. Moreover, some of the identified genes such as protein phosphorylation play a key role in most cellular activities. Additionally they serve as a reversible process mediated by protein kinases and phosphoprotein phosphatases (Hanks and Quinn, 1991). Protein kinases catalyse the transfer of the gamma phosphate from nucleotide triphosphates (often ATP) to one or more amino acid residues in a protein substrate side chain, resulting in a conformational change affecting protein function (Hanks and Quinn, 1991). In contrast to the findings of Agre et al (Agre et al., 2021), they identified single nucleotide polymorphisms (SNPs) on chromosome 3 that are linked to yam mosaic virus. These SNPs are located close proximity to genes encoding the AP2/ERF domain, AUX/IAA protein, major facilitator, and sugar transporter-like proteins. The findings of these studies might expedite the use of SNP variations to assist selection decisions in breeding white Guinea yam for genotype selection with resistance to mosaic virus in Uganda.

For the yam tuber yield we identified several important genes related to total tuber weight, including Phoenix dactylifera coatomer subunit delta-2-like, COMPASS-like H3K4 histone methylase component WDR5AARATH, Vicilin-like seed storage protein At2g28490ARATH, Ananas comosus IAA-amino acid hydrolase ILR1-like 3 and Late exocytosis, associated with Golgi transport. According to Li and Roberts (2001), the Vicilin-like seed storage protein At2g28490 protein family is a representation of the conserved barrel domain of the 'cupin' superfamily. This family contains 11S and 7S plant seed storage proteins, and germins. Plant seed storage proteins provide the major nitrogen source for the developing plant (Dunwell, 1998). Also, Ananas comosus IAA-amino acid hydrolase gene consisted of 4 beta strands and two alpha helices which make up the dimerization surface of members of the M20 family of peptidases. This family includes a range of zinc metallopeptidases belonging to several families in the peptidase classification. Family M20 are Glutamate carboxypeptidases. Peptidase family M25 contains X-His dipeptidases (Rowsell et al., 1997).

The identification of protein function led to the discovery of two putative genes related to dry matter content. The potential genes identified on chromosome 18 including Transcription initiation factor TFIID subunit 12b and Growth-regulating factor 1 at 24.1 Mb and 21.6 Mb respectively. The TFIID is one of several General Transcription Factors (GTFs), which also include TFIIA, TFIIB, TFIIE, TFIIF and TFIIH, that are involved in the accurate initiation of transcription by RNA polymerase II in eukaryotes (Gazit et al., 2009). TFIID plays an important role in the recognition of promoter DNA and assembly of the pre-initiation complex (Gazit et al., 2009). In addition, Kim (Kim et al., 2003) indicated that the WRC domain, named after the conserved Trp-Arg-Cys motif, contains two distinctive features: a putative nuclear localization signal and a zinc-finger motif (C3H). It is suggested that the WRC domain functions in DNA binding. In their study on water yams, (Gatarira et al., 2020) presented findings that found QTLs correlated with tuber dry matter at different sites. (Kayondo et al., 2018) discovered similar results and revealed that traits such as panel size, harvest time, and environment influence the discovery of quantitative trait loci (QTLs) related to virus disease. To address this issue, the most effective approach for enhancing the connections between traits and (QTLs) is to combine phenotypic scores from numerous locations with genotypic data utilizing diverse panels of individuals (Jing et al., 2018).

The study revealed that loci with varying effects can influence the variability of the traits examined in *Dioscorea rotundata*. This study identified distinct genomic regions harboring genes related to flowering development and disease resistance in yam germplasm. These findings should be further confirmed and evaluated. To do this, the QTLs may be converted into cost-effective Kompetitive Allele-Specific PCR (KASPs) markers, which can then be efficiently utilized to transfer alleles into high-quality yam genotypes. The results of this work might potentially aid in the development of novel breeding strategies for preserving advantageous genetic traits related to disease resistance and tuber production in certain yam genotypes, with the aim of enhancing future marker-based breeding efforts. The chromosomal regions responsible for these analyzed traits might be utilized for the purpose of selecting and efficiently combining advantageous alleles in order to enhance the population of *Dioscorea rotundata*. Certain roles and characteristics of the identified genes remain unexplored. Future studies on investigating the action mode of these genes in yam will help to elucidate the expression/regulation of these

4 Conclusion

Using five models, we identified 12 SNP markers associated with the three important traits. Through the allele's analysis, we identified as well promising alleles to be used for markers-assisted selection. Several genes linked to plant defense mechanism, plant growth, to tuber accumulation were reported. Future investigation such as gene expression may be required for gene profiling to understand the genetic basis in Uganda's *D. rotundata* gene pool. Moreover, the discoveries might prove valuable for the verification and integration of markers in the process of yam breeding.

Data availability statement

The original contributions presented in the study are publicly available. This data can be found here: FigShare, https://figshare.com/articles/dataset/good_vcf_uganda_emma_vcf/24957663?file=43950204.

Author contributions

EA: Conceptualization, Data curation, Formal analysis, Funding acquisition, Investigation, Methodology, Project administration, Resources, Software, Supervision, Validation, Visualization, Writing – original draft, Writing – review & editing. TO: Conceptualization, Supervision, Validation, Visualization, Writing – review & editing. WE: Conceptualization, Supervision, Validation, Visualization, Writing – review & editing. RB: Validation, Visualization, Writing – review & editing. PAr: Formal analysis, Software, Validation, Visualization, Writing – review & editing. PAd: Resources, Visualization, Writing – review & editing. EC:

Visualization, Writing – review & editing. AA: Formal analysis, Validation, Visualization, Writing – review & editing. ID: Validation, Visualization, Writing – review & editing. SM: Visualization, Writing – review & editing. RE: Funding acquisition, Visualization, Writing – review & editing. AO: Validation, Visualization, Writing – review & editing. MO-S: Visualization, Writing – review & editing. TA: Funding acquisition, Visualization, Writing – review & editing.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fhort.2024.1365567/full#supplementary-material

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