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Genetic diversity, population structure and marker-trait associations in Indian kale (*Brassica oleracea* L. gp. acephala) using cross-species microsatellite markers

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ABSTRACT

Kale is known for its exceptional nourishing and functional benefits to human body. However, it is an understudied species from genomic as well as agronomic aspects. It is important to characterize niche kale germplasms around the world to systematically conserve and utilize its genetic variability, especially for commercial traits in the interest of growers, consumers and industry. With this view, genomic and phenotypic characterizations of 62 Kashmiri kale accessions including popular landraces were done to estimate and partition genetic diversity, understand trait relationships, develop population structure and divulge marker-trait associations of economic significance. Sixty-six cross species microsatellite (SSR) markers within Brassica genus amplified 269 alleles in the germplasm. Their polymorphic information content (PIC) ranged from 0.00078 to 0.953 with an average of 0.407. The population structure analysis and neighbour joining tree clustering categorized the germplasm into three sub-populations. AMOVA revealed more within-population variance (67.73 %) than among-populations (32.27 %) variance. The principal component analysis (PCA) involving 24 agronomical traits revealed seven PCs (PC1 to PC7) having Eigen values more than 1, which explained a cumulative variation of 69.21 %. Association mapping with respect to these 24 agronomical traits using mixed linear model and general linear model revealed six overlapping significant marker-trait relationships with five being significant at probability value of 0.001/0.0001. The highly significant associations of two SSRs with economically important traits (siliqua length and seed weight) significantly correlated/ related with leaf yield and seed yield were revealed for their possible utilization in marker assisted breeding for higher leaf and seed yields.

1. Introduction

Referred to as the dog of the plant world, Brassica oleracea L. oleracea is the most diverse species in plant kingdom just as dog is the

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most diverse animal species [1]. Native to South and West Europe and thriving on sea cliffs, *Brassica oleracea* L. *oleracea* is commonly called wild cabbage and is the progenitor of diverse morphotypes that rule the world-wide vegetable fields and gardens today. It has been domesticated into various cultivars, the best known among which are cauliflower (Botrytis group), cabbage (Capitata group), broccoli (Italica group), Brussels' sprouts (Gemmifera group), kohlrabi (Gongyloides group) and kale & collards (Acephala group). Also known as colworts by the 19th century English, the Acephala (without head) group also became favourite of English after the

Germplasm of kale (Brassica oleracea L. Gp. Acephala) used in the study.

Accession name	Number assigned	Origin	Identity
CITH-KC-1/17	1	Kashmir	IC-0650676
CITH-KC-1/21	2	Kashmir	IC-0650674
CITH-KC-3	3	Kashmir	IC-0650702
CITH-KC-9	4	Kashmir	IC-0650678
CITH-KC-14	5	Kashmir	IC-0650680
CITH-KC-18	6	Kashmir	IC-594205
CITH-KC-21	7	Kashmir	Collection
CITH-KC-26	8	Kashmir	IC-0650694
CITH-KC-26R	9	Kashmir	Selection
CITH-KC-40	10	Kashmir	IC 594203
CITH-KC-42	11	Kashmir	IC-0650701
CITH-KC-44	12	Kashmir	IC 594207
CITH-KC-48	13	Kashmir	IC-0650683
NW-Sag-36R	14	Kashmir	IC-0650681
NW-Sag-21	15	Kashmir	IC-0650686
Kawdari	16	Kashmir	Landrace (IC-0650671)
NW-Sag-27	17	Kashmir	IC-0650670
CITH-KC-24KK	18	Kashmir	Selection
CITH-KC-38	19	Kashmir	IC-0650673
NW-Sag-41	20	Kashmir	IC-0650669
Pusa-Sag-1	21	Himachal Pradesh	Collection
NW-Sag-1	22	Kashmir	IC-0650675
Pusa-Sag-2	23	Kashmir	Selection
NW-Sag-29	24	Kashmir	IC-0650/13
HW-I	25	Kashmir	IC-0650677
Pusa-Sag-3	26	Kashmir	Collection
GM Duri	27	Kashmir	IC-0650682 (Landrace)
CITH-RC-SEI-S Vhanvari (dworf)	20	Kashmir	IC-0030/11
NW Sog 30	30	Kashmir	IC 0650687
CITH KC 4/21	21	Kashmir	IC 0650672
NW-Sag-38	32	Kashmir	IC-0650685
NW-Sag-44	33	Kashmir	IC-0650689
NW-Sag-40	34	Kashmir	IC-0650688
NW-Sag-24	35	Kashmir	IC-0650684
NW-Sag-15	36	Kashmir	IC-0650692
Hanz haaa	37	Landrace	IC-0650693 (Landrace)
NW-Sag-30	38	Kashmir	IC-0650691
Siberian	39	Exotic	Collection from private farm
CITH-KC-2	40	Kashmir	IC-0650695
Khanyari	41	Kashmir	IC-0650697 (Landrace)
CITH-KC-24	42	Kashmir	IC 594206
CITH-KC-Sel-5	43	Kashmir	IC-0650699
NW-Sag-33	44	Kashmir	IC-0650700
NW-Sag-23	45	Kashmir	IC-0650698
CITH-KC-4-Sel1/15	46	Kashmir	Selection
CITH-KC-16	47	Kashmir	IC-0650710
NW-Sag-42	48	Kashmir	IC-0650704
NW-Sag-4	49	Kashmir	IC-0650706
CITH-KC-7	50	Kashmir	IC 594208
CITH-KC-Sel-1	51	Kashmir	IC-0650705
CITH-KC-20	52	Kashmir	Collection
NW-5ag-20K	53	Kashmir Kashmir	IC-06506/9
	54	Kasiiniir	Collection
Japanese Green	55 E6	EXOLIC	Collection from private farm
	50	Na5IIIIII Kashmir	Collection
	59	Kashmir	
GIIII-KC-0/21 LIW 5	50	Kashmir	IC 0650709
CITH-KC-53	60	Kashmir	IC-0650709
Diisa-Sao-4	61	Kashmir	Selection
CITH-KC-3/21	62	Kashmir	IC-0650707

popularity of heading types (broccoli, cauliflower, cabbage) [2]. The present-day var. *viridis* of convar. *acephala* is known as collard and is the American derivative of English word 'colewort'. The variety *acephala* that too comes under convar. *acephala* is known as kale. Sometimes, the term kale is used collectively for both collard and kale. Kale (*Brassica oleracea* L. group Acephala), the king of leafy vegetables, with a century-old history, belongs to *Brassicaceae* family [3]. It is grown throughout the world with significant European areas, central and northern parts of Peninsula and parts of the Black Sea coast [4–7]. Owing to its nutraceutical and medicinal potential, kale has gained significant attention world-over and has been recognized as an exceptional source of phytochemicals. Kale, renowned as 'Superfood', with its health benefits and valuable nutritional composition, is a rich source of vitamin A, vitamin K, carotenoids, glucosinolates, lutein, zeaxanthin, iron, magnesium and protein [8–11]. It is loaded with powerful antioxidants like quercetin and kaempferol [12,13] and is a good source of folic acid, the pregnancy vitamin. Due to these properties, pharma/nutraceutical companies are now coming up with dietary supplement formulations prepared from kale [14]. Kale bread, kale juice, kale puree and kale microcapsules are some of the famous value-added products and formulations [15].

Nutraceutical and bioactive profiling is very important for crop improvement programs [16] and diversity assessment [17] of horticultural crops. Kale in India is primarily grown, traded and consumed in the Union Territory of Jammu and Kashmir, particularly in Kashmir region where it is a traditional crop. Probably, the most comprehensive known diversity of kale (*Brassica oleracea* gp. Acephala) in India is found in Kashmir valley embedded in Indian Himalayan Region. Kashmiri kale germplasm is morphologically highly diverse with several landraces markedly different from each other. Among many landraces, *Khanyari* is known for its succulent and puckered leaves, indeterminate columnar growth and adaptability to warmer temperatures, *GM Dari* bears large flat leaves, has rosette appearance and is highly adapted to cool season. *Hanz Haaq* is a short stature plant with rosette of tightly arranged leaves that keep appearing indefinitely ensuring multiple harvests. Since, kale is a highly cross-pollinated crop, it is common to find useful variants in farmers' fields. The accessions developed from these variant collections have also been found to have considerable phenotypic diversity. This diverse range of kale, whether among landraces or inbred accessions, can be effectively utilized for its genetic enhancement. However, it needs to be genetically characterized first.

Genetic characterization of any crop species germplasm is vital to utilizing desirable genes and developing their conservation strategies [18]. It helps carve population structures, elucidate genetic diversity, measure genetic relatedness and differences and understand evolution of populations and species on the whole. Nowadays, molecular markers play central role in achieving these goals. The most widely used molecular markers for the purpose are simple sequence repeats (SSR) or microsatellites, as they are co-dominant, repeatable, often polymorphic, and widely distributed throughout the genome [19].

Association mapping has become a routine practice in linking available markers with quantitative trait loci (QTLs) of importance in crop species [20]. Because association mapping has a higher mapping resolution than linkage mapping, it is frequently used to identify QTLs in crops [21]. Association mapping is based on linkage disequilibrium (LD) studies to evaluate statistically significant marker-trait associations (MTA). At the population level, association mapping takes advantage of past and evolutionary recombination. Other benefits include the ability to analyse several alleles at once and review numerous recombination events, in numerous generations, and in natural populations without creating a mapping population. Recently, MTA have been established using association mapping for quality and yield traits in crops such as tomato [22], maize [23], *Brassica napus* [20] watermelon [24], spinach [25] and *Brassica rapa* [26].

It has been hypothesized that landraces and naturally occurring variants in farmer fields are more diverse than modern bred cultivars and thus better sources of genetic variation for the purpose of improving a crop genetically [27,28]. Therefore, this study on Kashmiri kale landraces and inbred lines developed from farmer field collections was designed to estimate genetic diversity, attempt clustering of groups, develop population structure and partition the total variation into conveniently useable components for possible use in kale breeding and conservation. Secondly, the microsatellites or the SSRs are known to be highly conserved across species, more so within the same genus, and cross reproducible in species with no known SSRs. It was thus pertinent to study SSRs developed in other *Brassica* species for their reproducibility and practical utility in kale in the form of associations with economically important leaf and seed yield related traits besides other traits of interest.

2. Materials and methods

2.1. Germplasm

The germplasm has been developed from collections made from farmer fields across Kashmir valley and a few collected from other sources. Each accession is being maintained in active field gene bank at ICAR-Central Institute of Temperate Horticulture, Srinagar, Jammu and Kashmir in Karewa soils 33.5901°N and 74.4780 °E at 1500 m above mean sea level for more than 2–15 years. The accessions or germplasm lines are being inbred every year by caging 3 to 6 true-to-type plants together to circumvent cross-pollination with other lines. The details of the kale genotypes in presented in Table 1.

This germplasm was evaluated for 24 metric traits leaf blade length/LBL (cm), leaf blade diameter/LBD (cm), leaf blade number of incisions/LBNI, leaf blade thickness/LBT (mm), total number of leaves/NL, number of leaf lobes/NLO, petiole length/PeL (cm), petiole diameter/PeD (cm), vegetative stem length/VSL (cm), vegetative stem width/VSW (cm), leaf yield/LY (q/ha), initiation of flowering/ IF (days), end of flowering/EF (days), flowering period/FP (days), size of floral buds/SFB, floral stalk length/FSL (cm), number of floral buds per plant/NFBPP, days to maturity/DM, siliqua length/SL (cm), siliqua width/SW (cm), number of seeds per siliqua/NSS and 1000 seed weight/1000 SW (g), plant height/PH (cm) and canopy diameter/CD (cm) and 26 non-metric traits: plant shape, plant stature & main stem branching, leaf blade shape, leaf blade margin, leaf color, leaf angle (degree), leaf apex shape, leaf midrib curvature, leaf blade curvature, leaf midrib colour, leaf anthocyanin colouration, leaf undulation, leaf blade curling, leaf pubescence, stem

colour, petiole colour, inflorescence axis, floral stalk branching, floral bud color, flower color, flower size, silique attitude, silique surface, silique colour before drying and seed colour. Some of these traits have been shown in Fig. 1.

2.2. Molecular characterization

2.2.1. DNA extraction

The method described by Doyle and Doyle [29] was used to extract genomic DNA from 1g of young leaf tissue of different kale genotypes. The integrity and quality of genomic DNA was checked on 0.8 % agarose gel. Nano-drop spectrophotometer was also used to examine the amount and purity of DNA (ThermoScientific, Waltham, MA, USA). For subsequent use, high-quality DNA was diluted to a concentration of 25 ng/µl.

2.2.2. SSR amplification

After literature survey, 75 SSR markers were selected based on the reported polymorphic information content (PIC) values. The PCR were carried out in a thermal cycler (Takara, Japan) using 35 cycles of initial denaturation at 94 °C for 30 s, annealing at 45–58 °C for 30 s, extension at 72 °C for 1 min, and final extension at 72 °C for 10 min. The PCR products were resolved in a 3 % agarose (MolBio

Plant shape



Inverted pyramid (CITH-KC-18)



Column (*Kawdari*)



Prostrate (Pusa-Sag-1)



Dome (NW-Sag-40)



Shortened nonbranching stem supporting leafy rosette (CITH-KC-24)



Elongated branching stem supporting leaves (*Kawdari*)



Shortened branching stem supporting leaves (*Hanz haaq*)



Elongated nonbranching stem supporting leaves (*Khanyari*)



Extreme thickening (CITH-KC-48)



Intermediate thickening (NW-Saag-30)



Slight thickening (Khanyari)

Stem thickening

Plant stature and main stem branching

Fig. 1. Some non-metric traits expressed by the different kale accessions under study.

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Green (NW-Sag-40)

Petiole color



Purple (NW-Sag-36 Red)



(NW-Sag-27)



Purple (CITH-KC-8 Red)

Stem color



Lanceolate (NW-Sag-23)



Obovate (*Khanyari*) Leaf shape



Orbicular (*Kawdari*)



Lobed (Pusa Sag-1)



Entire (Kawdari)



Crenate (CITH-KC-16)



Dentate (NW-Sag-36R)



Serrate (CITH-KC-26)



Yellow (CITH-KC-53)



Green (NW-Sag-15) (F Floral bud color



Purple (Pusa-Sag-2)

Fig. 1. (continued).

HiMedia) gel at constant 120 V for 2.5 h, and the size of the amplicon was determined using a 50 bp DNA ladder (HiMedia). The bands were examined under UV light using the gel documentation unit (Bio-Rad Laboratories Inc., USA). During the course of study, nine markers amplified none of the DNA samples. So, eventually only 66 markers contributed to the study (Table 2). These SSRs belonged to the chromosomes of *Brassica napus*, *B. oleracea*, *B. rapa*, *B. nigra* and *Raphanus sativus*. Fig. 2(a and b) display bands that were amplified using the primers Ol10-G08 and Ol12-G04.

Table 2		
List of sixty-six microsatellite markers used in the st	udy along with their chromosomal location	, annealing temperature and primer sequence.

SSR no.	Marker	Chromosomal location	AT	Forward primer (5'–3')	Reverse primer (5'–3')
ssr1	Na10-E02	13	55	TCGCGCATGTAATCAAAATC	TGTGACGCATCCGATCATAC
ssr2	Na12-C08	11	55	GCAAACGATTTGTTTACCCG	CGTGTAGGGTGATCTAGATGGG
ssr3	Na14-C12	9	54	CACATTTTGGTTCAATTCGG	TACGACGCTGGTTTCGATTC
ssr4	Na14-D07	1	55	GCATAACGTCAGCGTCAAAC	CTGCGGGACACATAACTTTG
ssr5	Na12-C07	3	58	ACTCAACCCCACAAACCTG	AGTTCCCCGGATCCGATTAG
ssr6	Ol10-A05	2	55	TGTAATAACCCGACCCATCC	CTCTCTCGCTCTCTCGATCC
ssr7	Ol11-H02	4	55	TCTTCAGGGTTTCCAACGAC	AGGCTCCTTCATTTGATCCC
ssr8	Ol11-G11	13	54	GTTGCGGCGAAACAGAGAAG	GAGTAGGCGATCAAACCGAG
ssr9	Ol10-F11	11	55	TTTGGAACGTCCGTAGAAGG	CAGCTGACTTCGAAAGGTCC
ssr10	Ol10-H04	17	55	TCACCCCTCTATATCCACCC	CAGAATCTGCCTGAACATCG
ssr11	BRMS-040	1	55	TCGGATTTGCATGTTCCTGAT	CCGATACACAACCAGCCAACTC
ssr12	Ol10-H02	12	54	TCGGATTTGCATGTTCCTGACT	CCGATACACAACCAGCCAACTC
ssr13	Ni2-A11	16	55	AACAAACAAGAGTCGAATACGG	AATGCCCTCTAACTGAGCCC
ssr14	Na10–F06	3	55	CTCTTCGGTTCGATCCTCG	TTTTTAACAGGAACGGTGGC
ssr15	BRMA-019	5	55	CCCAAACGCTTTTGACACAT	GGCACAATCCACTCAGCTTT
ssr16	BRMS-008	3	55	AGGACACCAGGCACCATATA	CATTGTTGTCTTGGGAGAGC
ssr17	Ra2-A11	9	55	GACCTATTTTAATATGCTGTTTTACG	ACCTCACCGGAGAGAAATCC
ssr18	Ra2-E03	10	56	AGGTAGGCCCATCTCTCTCC	CCAAAACTTGCTCAAAACCC
ssr19	Ra2-E11	13	55	GGAGCCAGGAGAGAAGAAGG	CCCAAAACTTCCAAGAAAAGC
ssr20	Ra2-E12	8	55	TGTCAGTGTGTCCACTTCGC	AAGAGAAACCCAATAAAGTAGAACC
ssr21	0112-E03	7	55	CITGAAGAGCITCCGACACC	GACGGCTAACAGTGGTGGAC
ssr22	0112-F11	1	55	AAGGACICATCGIGCAATCC	GIGICAGIGGCIACAGAGAC
ssr23	0113-C12	8	55	AGAGGCCAACAAGAACACC	GAAGCAGCACCAGTGACAAG
ssr24	OIIO-FIIa	-	55	TITIGGAACGTCCGTAGAAGG	CAGCIGACIICGAAAGGICC
ssr25	Na14-B03	-	55	GAIGGICGCCGAIICAAIGA	CCCATCAGCACIAGAAACCA
ssr20	Na14-E08	14	55		
SSIZ/	0111 406	1/	55 EE		
ssr20	Na12 C06	19	55		TACCCCCTCTTATTCCATCC
ssr30	Na12-000	1	56	GGTAAGCCAAAAACCCTTCC	GAAACCGGTAACAAAGTCGG
ser31	Na12-D03	13	56	ACTGCCTACATGAGTTTCAGTG	GAGGGAAGACAACTGGTCTCA
ssr32	0110-G09	15	56	TGCTTCCTTTTTCTTCGCTC	GAAGCACGAACGCGAGAG
ssr33	Na12_F12	13	56	CGTTCTCACCTCCGATAAGC	TCCGATGTAGAATCAGCAGC
ssr34	O111-B05	3	55	TCGCGACGTTGTTTTGTTC	ACCATCTTCCTCGACCCTG
ssr35	O110-H07	16	55	TAGAGATGTCACCCGAAGGC	AGCTTCATTTCAGTCGGTGG
ssr36	Ol12-B05	14	55	GGAAAGCGAAGAGTGACGAC	ATTGGGTAAAGCTGTGCTCG
ssr37	Ni2-B01	19	55	AAGGAGATTGTTTTTGGGGGC	AAGACTAATAAACACACGGCG
ssr38	Ni4-E08	9	55	GATTTTGAGGAAGCGGAGG	CAAAGCACTGAGAGAGAGAGAGAGAG
ssr39	Ra2-F11	13	55	TGAAACTAGGGTTTCCAGCC	CTTCACCATGGTTTTGTCCC
ssr40	Ra2-G09	1	55	ACAGCAAGGATGTGTTGACG	GATGAGCCTCTGGTTCAAGC
ssr41	Ra2-E07	10	55	ATTGCTGAGATTGGCTCAGG	CCTACACTTGCGATCTTCACC
ssr42	Ol11-C02	17	55	GCATTGCAATCTTGTTGGTC	CGTTTCCATACAGATCGTAAGAC
ssr43	Ol13-C03	19	57	GATCGGAGATGCGATGAGAG	GCATGCACCAGTGAAAAACTC
ssr44	Ra2-D04	6	55	TGGATTCTCTTTACACACGCC	CAAACCAAAATGTGTGAAGCC
ssr45	Ra2-A10	6	55	CCAGTGTGTGTGTGTGTGTGTG	TTTAACAGATAGCGCAGTGGTC
ssr46	Ol12-G04	8	55	CGAACATCTTAGGCCGAATC	GGTTAACCTGCGGGATATTG
ssr47	Ra3-H10	5	55	TAATCGCGATCTGGATTCAC	ATCAGAACAGCGACGAGGTC
ssr48	Ra2-E04	-	55	ACACACAACAAACAGCTCGC	AACATCAAACCTCTCGACGG
ssr49	OI13-E08	12	55	TTCGCAACTCCTCCTAGAATC	AAGGTCTCACCACCGGAGTC
ssr50	Ra2-A01	7	55	TTCAAAGGATAAGGGCATCG	TCFTCFTCFTFTGFTGFCFTCCG
ssr51	Ra3-D02B	3	55	CACAGGAAACCGTGGCTAGA	AACCCAACCTCAACGTCTTG
ssr52	Na12-C03	12	55	ATCGTTGCCATTAGGAGTGG	ACCAAATTAACCCTCTTTGC
ssr53	Na12-H04	11	55	THATCGICHTICCCCICCC	ACAAGGAACTAGAGAGAGAGAGAG
ssr54	0112-D05	18	47		TAAGAGGGGACIICIAIIGGG
SST55	Raz-Huo	-	47		
ssr57	BRMS 005	5	17	ACCTCCTCCACATTCCTCTC	CCTCACCTTTCTTACCCCTC
ssr58	BRMS-036	1	47	GGTCCATTCCTTTTTGCATCTG	CATEGCAACGCCTAACAAACAT
ssr59	Na12_B09	-	47	ACGGAAGATCAAACAGCTCC	TGAGCGACCCATTCTTTAGG
ssr60	BRMS-015	_	47	TCGCCAATAGAACCCAAAACTT	CATCTCCATTGCTGCATCTGCT
ssr61	BRMS-027	1	47	GCAGGCGTTGCCTTTATGTA	TCGTTGGTCGGTCACTCCTT
ssr62	BRMS-042	_	47	AGCTCCCGACAGCAACAAAGA	TTCGCTTCCTTTTCTGGGAATG
ssr63	Ni4-B10	11	47	GTCCTTGAGAAACTCCACCG	CCGATCCCATTTCTAATCCC
ssr64	Na10-G10	-	45	TGGAAACATTGGTGTTAAGGC	CATAGATTCCATCTCAAATCCG
ssr65	Ol10-G08	14	54	TGCTTAATTGATTAGGGCAG	TTACCTCATCAGGTGGAGGC
ssr66	Ra2-A04	5	55	AAAAACTCCTCTTCAACG	CCCAAAGTTAGGTTTTAATGTAATCTC



(b)

Fig. 2. Documentation gel showing polymorphic bands of SSRs amplified by a) Primer Ol10-G08 b) Ol12-G04 (K1 to K31 are thirty-one accessions of kale).

2.3. Statistical analysis

2.3.1. Morphological

The datasets from two years were pooled. Statistical Tool for Agricultural Research (STAR) software was used to analyse the morphological data for descriptive statistics and principal component analysis (PCA). Program R [30] was used to determine the Pearson correlation coefficients for 24 agronomic traits.

2.3.2. Molecular

The SSR genotypic data was processed by PowerMarker v3.0 [31] to provide fundamental marker statistics values of expected heterozygosity (H_E), average heterozygosity (H_{av}), effective multiplex ratio (EMR), marker index (MI), number of alleles per locus, polymorphic information content (PIC), discriminating power (D) and resolving power (R). Using DARwin software, the unweighted neighbour joining tree was constructed based on genetic distances [32]. Program Arlequin v 3.5 [33] was used to estimate analysis of molecular variance (AMOVA).

For population structuring, model-based tool STRUCTURE 2.2 [34] was used. Using the admixture model with a burn-in of 10^5 and a run length of 10^5 the STRUCTURE programme was run five times for each K-value, ranging from 1 to 10. For each K, an average likelihood value [LnP (D)] was determined across all runs. In order to account for STRUCTURE's overestimation of the number of sub-groups, the most probable K value was determined using the *ad hoc* criterion (ΔK) of Evanno [35] estimated by STRUCTURE HARVESTER [36]. The accessions with membership probabilities of ≥ 0.80 were assigned to subgroups using the run of estimated numbers of subgroups displaying maximum likelihood. The admixed group received accessions with membership probabilities of < 0.80 [37].

Association mapping was carried out with TASSEL 3.0 [38]. To confirm the link between markers and traits, we employed two models-the mixed linear model (Q + K MLM) and the general linear model (Q GLM). The population structure (Q) matrix obtained from the STRUCTURE programme was used to perform the Q GLM approach. Ten thousand permutations were done to get a P-value between 0.05 and 0.01 for marker significance. The population structure Q matrix and kinship (K) matrix were employed in the Q + K MLM approach at P < 0.05 and P < 0.01. The kinship coefficient was calculated in SPAGeDi software [39]. The loci were considered to be in significant LD if P < 0.01.

3. Results

3.1. Morphological diversity and correlation

The analysis of variances conducted with respect to twenty-four morphological characteristics of the germplasm revealed

		0		0.01									
Source of Variation	DF	PH	CD	LBL	LBW	LBNI	LBT	NL	NLO	PeL	PeD	VSL	VSW
Replications	1	2.66	0.14	3.63	0.17	0.68	0.0	974.40	0.03	4.46	0.59	0.17	12.62
Genotypes	61	187.30*	74.79*	11.96*	8.82*	245.30*	0.02*	567.88*	7.14*	19.16*	4.79*	298.56*	558.31*
Error	61	2.04	3.10	0.67	0.43	1.56	0.001	1.53	0.01	0.74	0.22	3.80	8.99
Source of Variation	DF	SFB	FSL	NFBPP	IF	EF	FP	SL	SW	NSS	DM	SW	LY
Replications	1	0.06	0.007	79.04	3.23	44.16	20.16	1.11	0.02	6.78	28.07	0.30	16.48
Genotypes	61	3.69*	0.24*	2378439.5*	47.49*	103.45*	126.55*	2.41*	1.16*	35.27*	43.23*	2.50*	145.96*
Error	61	0.02	0.005	24.71	3.55	4.06	6.58	0.05	0.04	0.94	3.28	0.023	6.33

 Table 3

 Analysis of variance for various agronomical traits in 62 kale genotypes.

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**P < 0.01, *df* degree of freedom, *PH* Plant height, *CD* Canopy diameter, *LBL* Leaf blade length, *LBD* Leaf blade diameter, *LBNI* Leaf blade no. of incisions, *LBT* Leaf blade thickness, *NL* No. of leaves, *Nlo* No. of lobes, *Pel* Petiole length, *Ped* Petiole diameter, *Vsl* Vegetative stem length, *Vsw* Vegetative stem width, *SFB* Size of floral buds, *Fsl* Floral stalk length, Nfbpp no. of floral buds/plant,IF Initiation of flowering, EF End of flowering, FP Flowering period, SL Siliqua length, SW siliqua width, NSS No. Of seeds/siliqua, DM Days to maturity, SW Seed weight, LY Leaf yield. Values in the cells are mean sum of squares.

significant mean squares across the accessions (Table 3). The coefficient of variation (CV) for different traits ranged from 0.29 to 17.05 (Table 4). Number of leaf lobes (17.05 %) had the largest CV, followed by leaf yield (11.01 %), vegetative stem length (8.63 %), leaf blade thickness (8.41 %), flowering period (7.49 %) and petiole length (7.40 %). Leaf yield being complex quantitative trait showed strongest correlations with leaf blade diameter (0.347**), petiole length (0.247**) and siliqua width (0.364**) at P < 0.01. The findings of correlation analyses are shown in Table 5.

3.1.1. Morphological cluster analysis

According to the results of the cluster analysis of different traits, all accessions were divided into three clusters, with cluster I, III and II having 49, 8 and 5 accessions, respectively (Fig. 3). The y-axis of the dendrogram represents distance between clusters.

3.1.2. Principal component analysis

The PCA showed that eigenvalues for all PCs ranged from less than 1 to 4 (Table 6). Based on Kaiser criterion, seven PCs (PC1 to PC7) were found to have eigenvalues >1, which explained a cumulative variation of 69.21 % (Table 6). However, the Scree curve tends to become linear after PC3 (Fig. 4), we considered only first three PCs which explain the cumulative variance of 42.31 % (Table 6).

The PC1 explained the highest variation of 17.04 %, followed by PC2 explaining 15.25 % and then PC3 explaining 10.02 % of variation. Thus, the selection of lines and traits based on PC1 shall be most useful. The PC1 has positive association with 17 traits *viz.*, PH, CD, LBL, LBD, PeL, PeD, VSW, SFB, FSL, NFBPP, IF, SL, SW, DM, NSS, X1000SW, and LY whereas, negatively associated with 7 traits viz., LBT, LBNI, NL, NLO, VSL, EF, and FP (Table 7).

3.1.3. Molecular genetic diversity

Out of 75 SSR markers, nine markers didn't amplify any product in any genotype and were excluded from analysis. Sixty-six SSR markers amplified total 269 alleles on 62 kale accessions. The average number of alleles per locus was 4.07, with the range being 1 to 11. Marker Ol12-G04, showed the highest number of alleles, whereas a group of markers *viz.*, Na12–C07, BRMS-040, Ni2-A11, Ol10-G09, Ol11-C02, Ra2-H06, BRMS-005 and Na12–B09 amplified only one allele. Expected heterozygosity (H_E) ranged from 0.00 (BRMS-019, Ra2-F11, Ol12-D05, BRMS-042 and Ra2-H06) to 1.887 (Ol13-C12) with an average of 0.392. Similarly, effective multiplex ratio (EMR), average heterozygosity (H_{av}), marker index (M), discriminatory power (D) and resolving power (R) ranged from 0.0007 (Na12–C06) to 3.403 (Ra2-E07, Na10-G10), 0.00 (BRMS-019, Ra2-F11, Ol12-D05, BRMS-042 and Ra2-H06) to 2.483 (Ol10-G09), 0.00 (Ra2-F11, Ra2-F11, BRMS-042, Ol12-D05, Ra2-H06) to 2.451 (Ol10-G09), 0.00 (Ra2-F11, BRMS-019, BRMS-042, Ol12-D05, Ra2-H06) to 2.451 (Ol10-G09), 0.00 (Ra2-F11, Na14-E08, Na12–C06, Na12-D03, Na14-E02, Ol10-G09, Ra2-D04 and Ra2-A10) to 3.484 (Ol10-A05). The PIC value ranged from 0.00078 (Na12–C06) to 0.95348 (Na14-E02) and averaged to 0.407. Ten SSR loci *viz.*, (Ol13-C12, Ol11-H06, Na14-E02, Ra2-F11, Ra2-D04, Ra2-A10, RA3-D02B, Na12–C03, Ra2-E04, Ra2-A01 had PIC values greater than 0.5, demonstrating high level of discriminating potential of the selected SSRs (Table 8).

3.1.4. Population structure and molecular cluster analysis

A model-based application called STRUCTURE was used to divide the accessions into proper subgroups in order to study genetic relationships among them. The largest ΔK value was apparent for all lines when K = 3 (Fig. 5). Accessions were separated into three

 Table 4

 Descriptive statistics of twenty-four agronomic traits across 62 kale accessions.

S. No.	Trait	$Mean \pm SD$	Range	Coefficient of Variation (%)
1	Plant height (cm)	50.56 ± 9.69	20.80-90.60	2.82
2	Canopy diameter (cm)	53.34 ± 6.22	27.86-67.24	3.30
3	Leaf blade length (cm)	20.10 ± 2.51	14-29.20	4.10
4	Leaf blade width (cm)	14.41 ± 2.14	9.60-22	4.54
5	Leaf blade no. of incisions	19.27 ± 11.07	0.00-60	6.48
6	Leaf blade thickness (cm)	0.36 ± 0.1036	0.15-0.81	8.41
7	Number of leaves	30.52 ± 17.04	13.20-131.20	4.05
8	Number of leaf lobes	0.65 ± 1.88	0.00-7.40	17.05
9	Petiole length (cm)	11.65 ± 3.15	5.80-25.60	7.40
10	Petiole diameter (cm)	10.57 ± 1.58	5.73-13.77	4.48
11	Vegetative stem length (cm)	22.59 ± 12.25	13.80-87.80	8.63
12	Vegetative stem width (cm)	64.49 ± 16.78	20.06-93.71	4.65
13	Size of floral buds (mm)	$\textbf{7.45} \pm \textbf{1.36}$	4.48-11.12	2.18
14	Floral stalk length (cm)	1.57 ± 0.35	1–2.77	4.68
15	No. of buds/plant	1737.14 ± 1086.08	299–4116	0.29
16	Time of flowering (days)	252.37 ± 5.03	235–267	0.74
17	End of flowering (days)	$\textbf{286.45} \pm \textbf{7.33}$	274–302	0.70
18	Flowering period (days)	34.26 ± 8.14	15-60	7.49
19	Siliqua length (cm)	5.67 ± 1.11	2.83-9.37	4.15
20	Siliqua width (cm)	3.13 ± 0.77	1.40-4.82	6.49
21	No. of seeds/siliqua	16.90 ± 4.24	8.20-29.40	5.74
22	Days to maturity	321.85 ± 4.83	311-331	0.56
23	1000 seed weight (g)	$\textbf{4.71} \pm \textbf{1.12}$	2.10-7.80	3.21
24	Leaf yield (t/ha)	$\textbf{22.86} \pm \textbf{8.70}$	11.13-52.94	11.01

	0.00(+	0.200		0.2			0.257			0.22	0	0.000++
IF	-0.206*						0.251**		0.107*			0.233**
EF									0.197*			
FP).203*				0.20	2*	
SL				0.339	9**				-0.405**			0.298**
SW		0.191*		0.232	2**	-0.291**	-0.221*			0.27	1**	
NSS							-0.206*		-0.369**			0.246**
DM						-0.360**		-0.241**	-0.436**			0.201*
WS1000	0.519**			0.220	0*	-0.334**			-0.592**			0.231**
LY			0.210*	0.347	7**					0.24	7**	
	VSW	SFB	FSL	NFBPP	IF	EF	FP	SL	SW	NSS	DM	WS1000
РН												
PD												
LBL												
LBD												
LBNI												
LBT												
NI.												
NLO												
Pel.												
PeD												
VSI												
VSW												
SEB	0 200**											
ESI	0.299	0.191*										
NEPDD	0.200*	0.161										
IF	0.209"	0.305										
IF FF												
EF				0.00/*	0 450++	0.000						
FP		0.070++	0.07.1++	0.226*	-0.4/9^^	0.776	^					
SL		0.273**	0.274**		0.186*							
SW												
NSS				0.227*				0.394**				
DM					0.228*	0.217		0.317**				
WS1000								0.340**	0.238**		0.470*	*
LY								0.187**	0.364**			

	PH	CD	LBL	LBD	LBNI	LBT	NL	NLO	PeL	PeD	VSL
PH											
CD	-0.373**										
LBL		0.201*									
LBD			0.555**								
LBNI	-0.256**		0.258**								
LBT		-0.285^{**}									
NL		-0.388**	-0.378**	-0.304**		0.229*					
NLO	-0.535**			-0.306**	0.413**		0.244**				
PeL			0.362**		0.297**			0.245**			
PeD		0.271**	0.273**	0.273**			-0.517**	-0.341**			
VSL	0.543**	-0.570**	-0.373**		-0.368**	0.295**	0.414**		-0.311**	-0.225*	
VSW	-0.226*	0.368**	0.298**	0.258**		-0.203*	-0.430**	-0.224*		0.209*	-0.621**
SFB							-0.193*	-0.421**			
FSL	-0.275^{**}	0.241**									-0.245**
NFBPP		0.266**		-0.294**		-0.297**			0.223*		-0.239**
IF	-0.206*					0.251**				0.233**	
EF								0.197*			
FP					0.203*				0.202*		
SL				0.339**				-0.405**		0.298**	
SW		0.191*		0.232**	-0.291**	-0.221*			0.271**		
NSS						-0.206*		-0.369**		0.246**	
DM					-0.360**		-0.241**	-0.436**		0.201*	
WS1000	0.519**			0.220*	-0.334**			-0.592**		0.231**	0.177*
LY			0.210*	0.347**					0.247**		

Table 5Correlation coefficient for 24 traits of 62 kale accessions.

LY



Fig. 3. Dendrogram showing relationship among kale genotypes using Agglomerative clustering method.

Table 6
Summary of principal component analysis.

Statistics	PC1	PC2	PC3	PC4	PC5	PC6	PC7
Standard deviation	2.0226	1.9129	1.5506	1.3942	1.3446	1.1999	1.1245
Proportion of Variance	0.1704	0.1525	0.1002	0.081	0.0753	0.06	0.0527
Cumulative Proportion	0.1704	0.3229	0.4231	0.5041	0.5794	0.6394	0.6921
Eigenvalues	4.0907	3.659	2.4044	1.9438	1.8079	1.4398	1.2646



Fig. 4. Scree plot showing Eigen values for each principle component Fig. 4 is available only in black and white.

groups based on the membership probability criterion of 0.8 [40].

Sixty-two lines were assigned to the relevant 1–3 sub-populations using this method, accounting for 56.45 % (35), 9.68 % (6) and 27.42 % (17) of the total studied germplasm (Fig. 6). There were only four genotypes in the admixture group: CITH-KC-2, *Khanyari*, Japanese Green and NW-Sag-20R (Table 9). *Fst* mean values were found to be 0.4178 for sub-population I, 0.5892 for sub-population II (0.5892) and 0.3200 for sub-population III.

According to AMOVA, only 32.27 % genetic variance was detected among populations whereas 67.73 % was found within

PC3 0.2037 0.0986 -0.1239-0.2687-0.32990.0235 -0.0104-0.0820.0744 -0.04990.0138 0.0009 0.0981 -0.03680.4299 -0.34190.2974

0.4721

0.0484

0.0357

0.1723

0.1293

-0.2469

-0.0423

Eigenvectors of each t	trait for first three Principal Compone	ents.	
Traits	PC1	PC2	
PH	0.0204	-0.3634	
CD	0.2338	0.2733	
LBL	0.2569	0.1952	
LBD	0.283	-0.0655	
LBT	-0.165	-0.0796	
LBNI	-0.0714	0.3305	
NL	-0.3306	-0.0513	
NLO	-0.2733	0.3372	
PeL	0.0916	0.2265	
PeD	0.3148	-0.071	
VSL	-0.2585	-0.3653	
VSW	0.3021	0.1842	
SFB	0.2517	-0.023	
FSL	0.0899	0.1384	
NFBPP	0.0951	0.1019	
IF	0.0274	-0.1373	
EF	-0.0383	-0.056	
FP	-0.0919	0.0651	

0.2586

0.1695

0.1292

0.2212

0.1721

0.1868

Table 7
Eigenvectors of each trait for first three Principal Components

population (Table 10). A neighbour joining tree (Fig. 7) was constructed where the accessions were divided into three major groups using DARwin software, which produced results that were comparable to those of the STRUCTURE analysis and morphological analysis.

-0.1671

-0.073

-0.3055

-0.0574

-0.3293

0.026

3.1.5. Marker-trait associations (MTAs)

SL

SW

DM

NSS

LY

X1000SW

On the basis of combinations of the 66 SSR loci, an LD pattern was derived. D prime values ranged from 0.48 to 0.73, with an average of 0.58 and a significant LD value of 66.40 % was detected (Table 11). We identified 110 MTAs involving 30 SSR markers and 24 agronomic traits using a Q GLM model (Supplementary Table 1). There were six MTAs involving five markers using a Q + K MLM model (Supplementary Table 2). Information on significant MTAs overlap between Q GLM and Q + K MLM at a significance level of P \leq 0.01 (Table 12). In both Q GLM and Q + K MLM models, one marker ssr27 was found to be linked with plant height at significance level of P \leq 0.0001.

4. Discussion

For effective conservation, management, and exploitation of genetic resources in varietal development initiatives, a thorough characterization of plant genetic resources and knowledge of the genetic relationships in the germplasm collections are of prime importance. For optimal conservation, management, and utilization in varietal development, it is essential to characterize the germplasm and comprehend genetic relationships among genotypes. Such genetically characterized germplasms then act as genetic stocks for multi-target breeding programmes. The DNA markers further help in genetic improvement *via* selecting the appropriate members of a population on the basis of their association with the desired trait, technically called marker-trait association. Therefore, combining conventional and molecular approaches provides a better insight and enhances the credibility of morphological findings and interpretations to be utilized for the breeding of any crop.

The present study comprises the phenotypic as well as molecular characterization of indigenous kale accessions developed from farmer field collections that are expected to express considerable variability. Furthermore, microsatellite or SSR markers are deemed highly reproducible DNA markers across species. Therefore, in the lack of SSR in kale, those developed in other *Brassica* species were used in the study to ascertain their usefulness in achieving the above-mentioned objectives.

4.1. Genetic variability and relationships

In this study, we evaluated 62 accessions of kale indigenous to Kashmir, India. Significant differences were found for most of the morphological parameters. Leaf yield and number of lobes were found to have larger than 10 % CV, indicating the presence of significant genetic variation in the germplasm for the traits. The high CV in the characteristics is most likely caused by the fact that genotypes were collected from various local diversity hotspots. A large CV for leaf yield (11 % of 22.86 t/ha) is highly desirable, as it offers greater opportunity to improve this most important trait from growers' point of view.

Distribution of accessions among clusters on the basis of morphology grouped largest number of accessions into cluster I, which was followed by clusters III and II accounting for 79.03 %, 12.90 %, and 8.06 % of total germplasm. The clustering suggests that these kale

Genetic diversity parameters based on 66 SSR markers in 62 kale accessions.

Marker	Expected	Average	Effective	Marker	Discriminatory	Resolving	Polymorphism	No. of
	heterozygosity	heterozygosity	multiplex	index	power (D)	power (R)	information	alleles
	(H_E)	(H. av)	ratio (EMR)	(<i>MI</i>)			content (PIC)	(Na)
Na10-F02	0 4437	0.0014	1 6613	0.0024	0.8903	1 3871	0 3549	5
Na12-C08	0.3798	0.0012	1.2742	0.0016	0.9357	1.5161	0.3813	5
Na14-C12	0 2482	0.0013	0.4355	0.0006	0.9796	0.8710	0.4226	3
Na14-D07	0.2248	0.0015	0.8710	0.0032	0.2433	0.2581	0.4281	3
Na14-D07	0.4694	0.0008	3 3871	0.0032	0.2433	1 2258	0.3432	1
O110-A05	0.3723	0.0006	2 2258	0.0020	0.0300	3 4839	0.3841	9
0111 H02	0.4565	0.0000	2.2230	0.0015	0.9392	3 3226	0.3402	0
0111-1102	0.4303	0.0010	0.1452	0.0020	0.0702	0.2003	0.3492	7
0110 E11	0.0921	0.0004	0.1452	0.0001	0.9979	0.2903	0.4491	2
0110-F11	0.4403	0.0010	2.2903	0.0023	0.6955	1,0069	0.3303	3
0110-H04	0.2988	0.0016	0.5484	0.0009	0.9674	1.0968	0.4088	/
BRMS-040	0.2706	0.0010	0.6452	0.0007	0.9745	1.2903	0.4168	1
0110-H02	0.3906	0.0016	1.0645	0.0017	0.9300	0.3226	0.3143	4
N12-ATT	0.4745	0.0077	0.6129	0.0047	0.6282	0.7742	0.3619	1
Na10–F06	0.4694	0.0025	1.8710	0.0047	0.6123	1.7419	0.3432	3
BRMA-	0.00	0.00	2.0000	0.00	0.00	0.00	0.4534	2
019								
BRMS-008	0.4655	0.0010	2.5806	0.0028	0.8646	1.4839	0.3450	7
Ra2-A11	0.4870	0.0078	0.4194	0.0033	0.8281	0.8387	0.3348	5
Ra2-E03	0.4547	0.0024	1.0484	0.0026	0.8791	0.8710	0.3500	3
Ra2-E11	0.3906	0.0015	1.0645	0.0017	0.9300	2.1290	0.3771	4
Ra2-E12	0.4979	0.0040	1.0645	0.0043	0.7187	0.3871	0.3294	2
0l12-E03	0.498	0.469	0.387	0.469	0.498	0.317	0.317	4
0l12-F11	0.304	0.318	0.353	0.318	0.304	0.378	0.378	8
Ol13-C12	1.887	1.500	1.048	1.129	1.887	1.581	0.581	4
Ol10-F11a	0.002	0.002	0.002	0.003	0.002	0.001	0.001	4
Na14–B03	0.004	0.003	0.002	0.003	0.004	0.001	0.001	3
Na14-E08	0.50000	0.49925	0.30934	0.37661	0.34248	0.00	0.33936	2
O110-B01	0.33989	0.34027	0.41705	0.39398	0 40625	0 46489	0 40731	7
O111-H06	1 00000	2 40323	1.33871	1.25806	1.09677	1.00000	0.51613	6
Na12_C06	0.00403	0.00161	0.00071	0.00121	0.00110	0.00	0.00078	5
Na12-000	0.00403	0.00101	0.00095	0.00153	0.00121	0.00	0.00070	5
No14 E02	0.00403	0.00387	0.00095	0.00133	0.00121	0.00	0.00119	5
0110 C00	0.75205	0.70979	0.90378	0.93730	0.93244	0.00	0.93346	1
Ne12 E12	0.45101	2.40307	2.29032	2.45101	2.00432	0.00	0.09077	1
Nal2-FIZ	0.4667	0.4667	1.00000	0.4667	0.8631	0.45161	0.48/1	3
OII1-B05	0.4683	0.4683	1.51613	0.4683	0.8607	1.09677	0.4864	8
OI10-H07	0.4370	0.4370	1.33871	0.4370	0.8971	2.29032	0.5005	5
OI12-B05	0.4480	0.4480	2.40323	0.4480	0.8889	2.48387	0.4957	4
Ni2-B01	0.4895	0.4895	1.25806	0.4895	0.8183	2.45161	0.4763	2
Ni4-E08	0.4936	0.4936	1.09677	0.4936	0.6914	2.06452	0.4742	2
Ra2-F11	0.00	0.00	1.00000	0.00	0.00	0.00	0.5960	5
Ra2-G09	0.4752	0.4752	1.14516	0.4752	0.6265	1.70968	0.4832	5
Ra2-E07	0.5000	0.5000	3.40323	0.5000	0.7476	2.41935	0.4711	3
Ol11-C02	0.4870	0.4870	1.17742	0.4870	0.8249	1.83871	0.4775	1
Ol13-C03	0.4520	0.4520	1.24194	0.4520	0.8816	0.48387	0.4939	4
Ra2-D04	0.2835	0.2835	2.00000	0.2835	0.9712	0.00	0.5559	4
Ra2-A10	0.3584	0.3584	0.498	0.3584	0.9460	0.00	0.5318	3
Ol12-G04	0.4594	0.4594	0.317	0.4594	0.8728	0.45161	0.4905	11
Ra3-H10	0.4480	0.4480	0.387	0.4480	0.8865	1.09677	0.4957	5
Ra2-E04	0.4209	0.4209	0.469	0.4209	0.9105	2.29032	0.5075	5
Ol13-E08	0.4917	0.4917	0.469	0.4917	0.8144	2.48387	0.4752	5
Ra2-A01	0.4332	0.4332	0.498	0.4332	0.9006	2.45161	0.5022	5
Ra3-D02B	0.3122	0.3122	0.317	0.3122	0.9651	2.06452	0.5473	4
Na12-C03	0.3314	0.3314	0.498	0.3314	0.9569	0.00002	0.5411	6
Na12-H04	0.4480	0.4480	0.317	0.4480	0.8865	0.45161	0.4957	3
Ol12-D05	0.00	0.00	2.00000	0.00	0.00	0.00	0.46489	2
Ba2-H06	0.00	0.00	2,00000	0.00	0.00	0.00	0 46489	1
BRMS-033	0 50000	0.00403	1 00000	0.00403	0.75203	0.45161	0.33989	3
BRMS_005	0.33936	0.00078	1 51613	0.00110	0.95348	1 09677	0.40731	1
BRMS 026	0 30934	0.00071	1 33871	0.00005	0.96378	2.00077	0.41705	3
Nala POO	0.0000	0.00071	2 40222	0.00095	0.76070	2.27032	0.34007	1
Na12-BU9	0.49920	0.00101	2.40323	0.00387	0.70979	2.4030/	0.3402/	1
DRIVIS-015	0.3/001	0.00121	1.25800	0.00153	0.93/30	2.45101	0.39398	3
DKM5-027	0.34248	0.00110	1.090//	0.00121	0.95244	2.06452	0.40025	3
BRMS-042	0.00	0.00	1.00000	0.00	0.00	0.00	0.46489	3
N14-B10	0.47202	0.00254	1.14516	0.00291	0.85557	1.70968	0.35349	5
Na10-G10	0.48887	0.00099	3.40323	0.00335	0.81953	2.41935	0.34540	4
Ol10-G08	0.36006	0.00116	1.17742	0.00137	0.94513	1.83871	0.40007	6
Ra2-A04	0.42817	0.00173	1.24194	0.00214	0.90447	0.48387	0.37323	2



Fig. 5. Graphical depiction of population structure (A vertical line indicating membership in subgroup 1 (red), 2 (green), and 3 (blue) is displayed for each genotype. Genotypes are organised into K = 3 clusters according to the predicted membership coefficients. Rate of change of the likelihood distribution calculated as Delta K). (For interpretation of the references to colour in this figure legend, the reader is referred to the Web version of this article.)





accessions have sufficient levels of genetic variation. However, the relationship observed using morphological traits alone may not reflect the underlying genetic diversity. Therefore, analysis using molecular markers was also done to comprehend genetic differences among accessions.

SSR markers based molecular characterization of these accessions was carried out by using neighbour clustering method (Fig. 7). Cluster I (red) consisted of majority of the Kashmiri collections (36) and landraces *GM Dari* and *Hanz Haaq*, cluster II (green) consisted of eight collections, mostly non-indigenous and a landrace *Kawdari*. Cluster III (blue) consisted of 18 accessions, all representing indigenous Kashmiri collections and a landrace *Khanyari*. Thus, it clearly shows that each subgroup differs from the other. The "green" group is the most diverse, as it contains mostly non-indigenoua accessions (Himachal Pradesh, Private farms). The "blue" group is the least diverse because all its members are local accessions (Fig. 7).

Agglomerative clustering (Fig. 3) of the germplasm on the basis of morphology also seems to be considerably similar to that of SSR based neighbour joining tree. Landraces *GM Dari* and *Hanz Haaq* can be seen together in the largest cluster containing most of the indigenous accessions as they are in red cluster of neighbours joining tree, which is also the largest and hosts mostly the local kashmiri accessions and a few exotics from Himachal Pradesh. Similarly, landrace *Khanyari* is in a separate cluster from other landraces in both agglomerative clustering and SSR based neighbour joining tree. With respect to the third landrace *Kawdari*, interestingly it belongs to the clusters that are almost identical in the two cases. Both cluster II of Agglomerative clustering method and cluster 'green' of SSR based tree are identical with the exception of only one accession. Such concordance between morphological clustering, molecular clustering and geographical origins of the accessions indicates the accuracy of selected SSRs in truthfully characterizing kale germplasm.

Several studies have successfully used microsatellite markers to find molecular genetic variation in *Brassica* crops [41–44]. In the present study, degree of variation among accessions was examined using the diversity indices Na, H.av, He, EMP, MI, DP, and RP. The level of variability and allele diversity are indicated by the total number of alleles in the population. Here, the numbers of alleles varied from one to eleven per locus, and in total 269 alleles were amplified. This is very large number of alleles given the fact that SSR used are from different species. Using 12 SSRs on 25 accessions of *B. oleracea* including kale, El-Esawi et al., 2016 [41] obtained only 47 alleles indicating that using larger number of *Brassica* SSR markers on *B oleracea* varieties, as done in our study, will be beneficial in case of

Table 9	
Probability based assignment of individuals to the	(K) sub populations

Genotype	K1	K2	K3	Sub-population assigned
G1	0.991	0.001	0.008	1
G2	0.995	0.001	0.004	1
G3	0.995	0.001	0.004	1
G4	0.997	0.001	0.002	1
G5	0.992	0.001	0.006	1
G6	0.995	0.001	0.004	1
G7	0.989	0.006	0.005	1
68	0.997	0.001	0.002	1
G9 G10	0.98	0.008	0.012	1
G11	0.985	0.004	0.003	1
G12	0.996	0.001	0.003	1
G13	0.992	0.001	0.008	1
G14	0.005	0.994	0.001	2
G15	0.994	0.002	0.004	1
G16	0.002	0.997	0.001	2
G17	0.995	0.001	0.005	1
G18	0.984	0.001	0.014	1
G19	0.981	0.001	0.018	1
G20	0.99	0.001	0.009	1
G21	0.001	0.998	0.001	2
G22	0.975	0.022	0.003	1
G23	0.996	0.001	0.003	1
G24	0.988	0.001	0.011	1
G25 C26	0.986	0.001	0.013	1
G20 C27	0.922	0.001	0.077	1
G27 G28	0.990	0.001	0.003	1
G20	0.986	0.001	0.012	1
G30	0.99	0.001	0.009	1
G31	0.968	0.001	0.031	1
G32	0.972	0.001	0.026	1
G33	0.949	0.001	0.05	1
G34	0.909	0.003	0.088	1
G35	0.986	0.004	0.01	1
G36	0.995	0.001	0.004	1
G37	0.987	0.005	0.008	1
G38	0.985	0.001	0.013	1
G39	0.026	0.964	0.01	2
G40	0.577	0.001	0.422	Admixture
G41	0.255	0.005	0.74	Admixture
G42	0.016	0.001	0.983	3
G43	0.012	0.001	0.987	3
G44	0.002	0.001	0.997	3
G46	0.026	0.001	0.888	с З
G47	0.020	0.001	0.373	3
G48	0.035	0.001	0.964	3
G49	0.005	0.001	0.994	3
G50	0.006	0.001	0.993	3
G51	0.008	0.001	0.99	3
G52	0.077	0.003	0.92	3
G53	0.013	0.496	0.492	Admixture
G54	0.007	0.003	0.99	3
G55	0.003	0.533	0.465	Admixture
G56	0.006	0.003	0.991	3
G57	0.007	0.004	0.989	3
G58	0.008	0.002	0.99	3
G59	0.002	0.823	0.174	2
G60	0.031	0.001	0.968	3
G61	0.002	0.99	0.009	2
G62	0.028	0.002	0.97	3

kale. The expected heterozygosity (He) differed from 0.00 to 1.887 with an average of 0.392, which was higher than the mean heterozygosity (0.266). The bulk of the reported homozygotes may actually be heterozygotes, where one allele is present but the other is absent. Similar to our results the value of average He of 0.33 has been reported by Pelc et al. [45]. SSR markers were, therefore, found to be more effective in assessing genetic diversity and generated mean values for EMR, MI, DP, and RP of 1.177, 0.244, 0.751, and 1.155, respectively [46]. PIC values ranged from 0.00078 to 0.953, with an average of 0.407, indicating the high applicability of these

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Analysis of molecular variance (AMOVA) for 62 kale accessions.

Population	Sources of variation	df	Sum of squares	Estimated variability	Percentage of variation (%)	P value
	Among populations Within population Total Fst	2 57 59 0.4178 0.5892 0.3200	385.805 1248.812 1634.61	10.43 ^a 21.90 ^b 32.34	32.27 67.73	<0.05 <0.05



Fig. 7. Neighbour joining tree clustering of 62 genotypes of kale based on SSR data using DARwin.

Table 11	
The r ² , D prime and LD statistics of 62 kale a	ccessions.

Chromosome	Mean of r ²	Mean of D prime	LD (%)
1	0.13	0.64	38.60
2	0.03	0.50	55.98
3	0.07	0.50	50.15
4	0.04	0.49	53.53
5	0.04	0.61	66.17
6	0.03	0.60	64.39
7	0.08	0.53	43.35
8	0.06	0.52	49.97
9	0.11	0.63	39.88
10	0.17	0.73	30.17
11	0.09	0.61	39.80
12	0.04	0.62	56.11
13	0.05	0.54	52.65
14	0.12	0.68	45.41
15	-	-	-
16	0.03	0.48	58.54
17	0.04	0.54	58.11
18	0.14	0.68	46.97
19	0.02	0.56	66.40

Details on the overlapping significant marker-trait association between two models.

Chromosome	Marker	Trait	GLM	MLM	Average P value
4	ssr7	Plant height	**	**	**
17	ssr27	Plant height	***	***	***
16	ssr13	Petiole diameter	***	**	**
11	ssr9	Flowering period	*	**	*
1	ssr29	Siliqua length	***	**	**
11	ssr9	Seed weight	***	**	**

markers and suggest that the majority of the markers allowed for a high level of polymorphism. Our results were in conformity with the earlier findings [7,47].

Clustering generated through DARwin software using unweighted neighbour joining method was found to be the most effective in the current study for the interpretation of findings for determining the phylogenetic relationships between accessions and the closest and most distant genotypes from one another. All 62 genotypes were grouped into three main clusters/sub-populations using both an unweighted neighbour joining-based dendrogram and population structure analysis. Cluster III consisted of all the introduced accessions and two indigenous accessions. In cluster II, only one landrace '*Khanyari*' was spotted, the remaining 17 were indigenous collections. All other landraces *Kawdari, Hanz haaq*, and *GM Dari*, despite being phenotypically strikingly different clustered under Cluster I along with rest of the indigenous collections. The results of the structure analysis shared a similar pattern of relationship with slight mixing of genotypes. Previous studies have also observed genetic clustering based on marker data. Singh et al. [48] used the neighbour joining approach to divide 87 *Brassica juncea* genotypes into 2 groups. Likewise, Pelc et al. [45] divided the collard (*Brassica oleracea* L, var, *viridis*) landraces into 3 clusters.

In population structure studies under the present investigation, population differentiation measurements (Fst) ranged from 0.32 (sub-population III) to 0.58 (sub-population II), which are relatively high and confirm the separation of all the sub-populations, El-Esawi et al. [41] used 12 SSRs and observed an Fst value of 0.110 for genetic differentiation in kale accessions. The fact that we used much larger number of markers in our study may be the reason why we were able to achieve higher Fst values. Alternative reason can be a relatively strong genetic difference between genotypes explored in this study. The AMOVA showed that genetic diversity within subpopulation was higher than among subpopulations. Gene flow from other varieties nearby such as by cross-pollination between fields and/or seed exchange between farmers across the valley may be the cause of higher genetic variation within the sub-populations of Kashmiri kale [7]. Nevertheless, gene flow between subpopulation is also considerable given that more than 32 per cent genetic variation is due to among population differences.

There is an apprehension that population structure analysis may produce false positive findings. Therefore, to overcome this problem, we also did association mapping analysis by using general linear model (GLM) and mixed linear model (MLM). The MLM is able to address the problem of false positives in marker-trait association by accounting for both kinship (K-matrix) and population structure (Q-matrix), whereas the GLM only considers (Q-matrix) population structure [49,50]. For 24 agronomic traits involving 30 markers, we found 110 marker-trait associations based on the Q GLM model. The Q + K MLM model led to the identification of six marker-trait associations involving five markers. The GLM approach confirmed each and every association that the MLM approach had shown to be statistically significant (P < 0.001). There have been numerous other association mapping studies using different sample sizes in Brassicaceae family [51,52]. LD levels in our study enabled us to identify marker-phenotype relationships in the inbred lines. We observed two markers associated with same trait and same marker associated with two traits. Specifically, Ol11-H02 and Ol10-B01, located on chromosome 4 and 17 of B. oleracea associated with same trait 'plant height'; Ol10-F11 located on chromosome 9 of B. oleracea has association with flowering period and seed weight; Ni2-A11 located on chromosome 16 of B. nigra associated with petiole diameter and Na12-C06 located on chromosome 1 of B. napus, associated with siliqua length, respectively. Several plant association studies for various phenotypic traits, including flowering time [26], leaf and plant architecture [21,25] and fruit quality [53] have been similarly reported. Markers Na12-C06 of B. napus and Ol10-F11 of B. oleracea found to be associated with siliqua length and seed weight, respectively, in our study are of tremendous value in breeding programmes aimed at high leaf productivity since these traits were also found to have significantly positive correlation with leaf yield. Similar is the recommendation for Ol10-F11 marker from B. oleracea with respect to seed yield, as seed weight associated with this marker directly contributes to seed yield. Leaf and seed productivity, being unarguably the foremost traits of agricultural and commercial importance may significantly benefit from using these markers in kale breeding programmes.

5. Conclusion

This is the first study on morphological and molecular diversity assessment, population structure and MTA for agronomically important traits of Indian kale. We detected significant polymorphism at both morphological and molecular levels. The morphological evaluation of twenty-four metric traits has revealed significant CV for all traits and AMOVA revealed significant genetic diversity within and among populations. High level of diversity in the studied germplasm panel showed its potential application in kale breeding programs. The PCA revealed 7 principal components that explained 69.21 % variance in the germplasm. Ten of the 66 *Brassica* SSR markers selected for this study have shown PIC values between 0.5 and 0.953 suggesting their high potential in assessing molecular genetic diversity in kale.

The molecular clustering analysis has divided the whole germplasm into 3 sub-populations and a group of 4 admixtures. One of the popular landraces of Kashmir '*Khanyari*' has placed itself in cluster II while rest of the three '*Kawdari*, '*GM Dari*' and '*Hanz haaq*' have gone together into a different cluster (cluster I). As Khanyari is phenotypically very similar to *Kawdari* and on the other hand *Kawdari*, *GM Dari* and *Hanz haaq* are completely different from each other (Fig. 8), this finding reinforces the importance of DNA markers in genetic diversity assessment of crop species. The DNA markers can assess the genetic diversity more precisely in case of kale, as the ambiguity caused by morphology has been put at bay.

Considerable similarity between morphological and molecular clustering of the germplasm along with their consonance with geographical origins of the accessions was also observed in this study. This suggests strong candidature of the studied SSRs in diversity studies of kale.

The study also identified 110 MTAs involving 30 SSRs and 24 traits of kale. Six of these markers showed overlapping significant MTAs under GLM and MLM for 5 different traits. One of these markers, Ol10-B01 was found to associate with plant height at $p \le 0.0001$. This marker, therefore, shows promise in marker-assisted breeding (MAB) for plant architecture manipulation. Two markers Ol10-F11 and Na12–C06 were found to be significantly associated with seed weight and siliqua length, respectively. Both these traits showed significant positive correlation with leaf yield. Secondly, seed weight can be a direct contributor to seed yield. Therefore, both leaf yield and seed yield of kale may be improved through MAB by utilizing these markers.

Data availability statement

The data generated in the study is included in the article and supplementary material.

Statements and declarations



Khanyari



Kawdari



Hanz Haaq

GM Dari

Fig. 8. Kashmiri kale landraces.

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CRediT authorship contribution statement

Geetika Malik: Writing – review & editing, Resources, Project administration, Methodology, Funding acquisition, Formal analysis, Conceptualization. Asma Jabeen: Writing – original draft, Formal analysis, Data curation. Javid Iqbal Mir: Resources, Methodology. Rafiq Ahmad Shah: Formal analysis. Mohd Abas Shah: Formal analysis. Vishal Dinkar: Writing – review & editing. Muneer Ahmad Sheikh: Resources. Ravinder Kumar: Formal analysis. Om Chand Sharma: Visualization, Project administration. Mahendra Kumar Verma: Writing – review & editing, Visualization.

Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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Appendix A. Supplementary data

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