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Research Article

Allele mining for the grain number gene *An-1* in rice (*Oryza sativa* L.)

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Abstract
Rice yield has attained a plateau and hence the enhancement of grain yield is indispensable to feed the growing population, which could be achieved by the identification of superior alleles in the existing germplasm. Any variation in the pleiotropic gene, *An-1* (yield gene) leads to enhanced grain number and grain size in rice. Hence, the gene was chosen for analyzing the allelic diversity/haplotype variation with 150 lines of 3K RG panel which revealed that, the gene *An-1* has 20 Single Nucleotide Polymorphisms and 10 INDELs encompassing both intronic and exonic regions. The genotypes were divided into four haplotypes in the combination of seven SNPs with the maximum number of genotypes in the first haplotype and the least number of genotypes in fourth haplotype. From the study, H1 was identified as a superior haplotype. The haplo-pheno analysis identified the superior donors *viz.,* SIGARDIS, GENIT and DAMNOEUB KAUN KHMOM harbouring superior haplotype combinations, which may be further utilized in haplotype-based breeding for the development of high yielding rice varieties.

Key words: rice, *An-1,* grain number, haplotype, SNPs, allelic diversity.

INTRODUCTION

Rice *(Oryza sativa L.)* is life for almost half of the global human population. The world faces the challenge of feeding 9.7 billion people by around 2050 (Source: United Nations Department of Public Information). Owing to an increase in the global population and a decrease in arable land, upsurging grain yield has become a priority. Rice yield has attained a plateau for the past one decade. Hence, enhancement of grain yield is the foremost objective in rice-breeding programs (Umadevi *et al*., 2019, Singh *et al*., 2020). The yield potential of the crop can be increased by altering the photosynthetic rate (C4 rice), identification of novel genes from wild, distant relatives/germplasm (allele mining), or by the creation of novel alleles through targeted mutagenesis. Grain yield in rice, which is a quantitative trait is governed by the components

viz., the number of panicles per plant, the number of grains per panicle, and grain weight (Xing and Zhang, 2010). Among these, grain number per panicle is a highly variable component that mainly depends on the length of the panicle, number of primary and secondary branches, and percentage of filled grains (Deshmukh *et al.,* 2010). The untapped novel alleles in primitive cultivars/landraces are of immense value for developing superior cultivars (Fess *et al.,* 2011). Exploring allelic variations/haplotype variations pave way for identifying suitable donors with the desired trait of interest that can be further employed in crop improvement programmes (Varshney *et al*., 2018).

Domestication of rice was achieved through selection for deletion mutant, but genetic improvement in rice depends

on the search of novel alleles and assembly of them in favourable genetic background. In recent days, several genes related to grain yield in rice have been identified. Some of the key notable genes were *An1, Gn1a, DEP1, DEP3, APO1, GS3, GW5, PROG1, LRK1, OsSPL14, EP3,* and *SP1* which modulate grain number per panicle directly or indirectly (Gouda *et al.,* 2019). Among them, *An-1* is a pleiotropic gene that regulates grain size, grain number, and awn development in rice. Transgenic studies have confirmed that the *An-1* gene positively regulates the extension of lemma (*i.e.* awn elongation) and negatively regulates the grain number per panicle in rice. Attempts have been made to increase the grain number per panicle by downregulating the *An-1* gene through RNAi technology. There is an increase in grain number per panicle by around 13.60 to 38.4 per cent in RNAi lines. Hence, the downregulation of the *An-1* gene leads to an upregulation of the *LOG* (*LONELY GUY*) gene, which in turn leads to an enhanced cytokinin level and meristematic activity, resulting in increased grain number per panicle and thus yield (Luo *et al.,* 2013). Though numerous studies have been conducted for yield enhancement in rice, the impact of alleles related to yield traits are not very much clear and hence it is a pre-requisite to survey the allelic diversity in the germplasm of rice (Gouda *et al*., 2020). Allele mining of yield-related traits has been attempted by very few research groups on genes like *Gn1a, GS312, DEP114, Ghd713,* and *sd115* (Vemireddy *et al*., 2019). This study was aimed at unravelling the allelic variations in *An-1* encompassing 150 genotypes from 3K RG panel, a core collection of 3000 resequenced rice accessions and harnessing its potential opportunities in future plant breeding programmes.

MATERIALS AND METHODS
A total of 150 genotypes belonging to the 3K RG panel (Li *et al*., 2014) were raised in augmented design at Paddy Breeding Station, Tamil Nadu Agricultural University during *Rabi*, 2020 (**Table 1**). Twenty-five days old seedlings were transplanted in the main field at a spacing of 20 x 20 cm and the recommended package of practices were carried out with adequate fertilizer application of 150:50:50 N:P: K kg/ha, respectively. The number of grains per panicle was recorded on three randomly selected competitive plants in each genotype.

Descriptive statistics including mean, median, range (minimum, maximum) and coefficient of variation (CV) for the whole population were estimated using Minitab 19 Statistical Software. Frequency distribution and box plot graphs were also constructed using Minitab 19 Statistical Software (Allen, 2019).

The allelic diversity for the *An-1* gene (LOC_Os04g28280) was carried out for the studied genotypes by downloading their sequence using the Rice-SNP seek database (http://snp-seek.irri.org/) encompassing the non-synonymous SNPs and INDELs. Nipponbare (*Oryza sativa* ssp.

japonica) sequence of *An-1* gene was taken as a reference one and aligned against the study sequences using BioEdit software (Hall, 2011).

The haplotype analysis was carried out by downloading the allelic variations of the *An-1* gene for the respective genotypes from the Rice-SNP seek database in PLINK format and later converted into haploview format using PLINK software (Jonathan, 2010). Haplotype groups and Linkage Disequilibrium blocks of the allelic variation were constructed using Haploview 4.2 software with parameters including HW p-value (Hardy-Weinberg p value) of 0.001, minimum genotype per cent of 75 and minimum minor allele frequency of 0.001 (Barrett *et al*., 2005). The significant differences between the constructed haplotype groups were proved by Dunnett's test using Minitab 19 Statistical Software (Allen, 2019).

RESULTS AND DISCUSSION
Any distinct variation in the gene, *An-1* (yield gene) leads to increased grain number and grain size in rice and hence this gene was chosen for assessing its allelic diversity.

The phenotypic data of the 150 genotypes were subjected to descriptive statistics, which delineates the simple measures of variability. Estimates for mean, median, range (minimum, maximum) and coefficient of variation were (**Table 2**). It was observed that the coefficient of variation was high *i.e.* 27.35 (> 20%) indicating that there is an ample amount of variation in the studied population and hence, the selection of suitable donors for the respective trait can be carried out with the available genotypes. From the box plot curve, it is understood that the range varies from 47 to 223 number of grains per panicle with an average of 113.34 grains per panicle (**Fig. 1**). The histogram implies that the population follows a normal distribution (**Fig. 2**).

Multiple Sequence Alignment of 150 genotypes with reference nucleotide sequence for *An-1* gene identified 20 SNPs and 10 INDELs in the both intronic and exonic region, which is given in **Table 3** along with their minor allele frequency (MAF). Both synonymous and non-synonymous SNPs were identified. Since the synonymous SNPs do not cause any alteration in the amino acid, the non-synonymous SNPs were enumerated further. Among the non-synonymous SNPs, SNP1 induces a base replacement of G→A at 16734725 positions of nucleotide sequence, causing R→W in the coded protein. SNP2 causes G→C substitution at 16734542 leading to R→G in the coded protein. SNP13 induces T→C base replacement at 16732870 positions, causing E→G substitution in the encoded protein. SNP15 causes T→G replacement of nucleotide sequence in 16732750 positions leading to T→P substitution of amino acid. SNP17 induces C→T transition at 16732732 positions causing G→D substitution in the coded protein.

Table 1. List of genotypes used in the present study with its respective haplotype groups

S. No.	Accessions	H	S. No.	Accessions	H	S. No.	Accessions	H	S. No.	Accessions	H	S. No.	Accessions	H
1	IRIS 313-9867	H1	31	IRIS 313-9604	H1	61	IRIS 313-10640	H1	91	IRIS 313-9924	H1	121	IRIS 313-9626	H1
2	IRIS 313-8699	H1	32	IRIS 313-11139	H1	62	IRIS 313-11998	H3	92	IRIS 313-12187	H1	122	IRIS 313-9522	H2
3	IRIS 313-10768	H1	33	IRIS 313-10334	H1	63	IRIS 313-8986	H3	93	IRIS 313-9862	H1	123	IRIS 313-10983	H1
4	IRIS 313-10497	H1	34	IRIS 313-8768	H1	64	IRIS 313-11841	H1	94	IRIS 313-8924	H1	124	IRIS 313-10226	H1
5	IRIS 313-10001	H1	35	IRIS 313-10234	H1	65	IRIS 313-12007	H1	95	IRIS 313-9611	H1	125	IRIS 313-9740	H1
6	IRIS 313-11790	H1	36	IRIS 313-10235	H1	66	IRIS 313-12046	H1	96	IRIS 313-10397	H1	126	IRIS 313-8946	H1
7	IRIS 313-10575	H1	37	IRIS 313-9249	H1	67	IRIS 313-9898	H1	97	IRIS 313-9391	H1	127	IRIS 313-9218	H1
8	IRIS 313-8850	H1	38	IRIS 313-10412	H1	68	IRIS 313-8930	H1	98	IRIS 313-8771	H1	128	IRIS 313-12287	H1
9	IRIS 313-8585	H1	39	IRIS 313-11982	H1	69	IRIS 313-8514	H1	99	IRIS 313-11521	H3	129	IRIS 313-9709	H1
10	IRIS 313-11423	H1	40	IRIS 313-9320	H1	70	IRIS 313-12194	H1	100	IRIS 313-11359	H1	130	IRIS 313-11927	H1
11	IRIS 313-8968	H1	41	IRIS 313-11638	H1	71	IRIS 313-9403	H1	101	IRIS 313-8690	H1	131	IRIS 313-11178	H1
12	IRIS 313-9696	H1	42	IRIS 313-9590	H1	72	IRIS 313-11798	H1	102	IRIS 313-9429	H1	132	IRIS 313-10220	H1
13	IRIS 313-9705	H1	43	IRIS 313-8414	H1	73	IRIS 313-7816	H1	103	IRIS 313-10045	H1	133	IRIS 313-8481	H1
14	IRIS 313-9492	H1	44	IRIS 313-8844	H1	74	IRIS 313-10718	H2	104	IRIS 313-9019	H1	134	IRIS 313-10606	H1
15	IRIS 313-8492	H1	45	IRIS 313-8435	H1	75	IRIS 313-11723	H1	105	IRIS 313-9634	H1	135	IRIS 313-12261	H1
16	IRIS 313-11398	H1	46	IRIS 313-9778	H2	76	IRIS 313-7807	H1	106	IRIS 313-11664	H1	136	IRIS 313-9372	H1
17	IRIS 313-11849	H1	47	IRIS 313-8232	H1	77	IRIS 313-11351	H1	107	IRIS 313-9424	H1	137	IRIS 313-11050	H1
18	IRIS 313-9551	H1	48	IRIS 313-8717	H1	78	IRIS 313-9606	H1	108	IRIS 313-10337	H1	138	IRIS 313-12012	H1
19	IRIS 313-8996	H1	49	IRIS 313-10652	H1	79	IRIS 313-10726	H1	109	IRIS 313-9351	H2	139	IRIS 313-11091	H1
20	IRIS 313-10374	H1	50	IRIS 313-8647	H1	80	IRIS 313-8405	H1	110	IRIS 313-11727	H1	140	IRIS 313-9182	H1
21	IRIS 313-8412	H1	51	IRIS 313-9572	H1	81	IRIS 313-10065	H1	111	IRIS 313-8252	H4	141	IRIS 313-10341	H1
22	IRIS 313-10775	H1	52	IRIS 313-11467	H1	82	IRIS 313-11710	H1	112	IRIS 313-9072	H1	142	IRIS 313-8982	H3
23	IRIS 313-12053	H1	53	IRIS 313-12078	H1	83	IRIS 313-9940	H1	113	IRIS 313-9555	H1	143	IRIS 313-9516	H4
24	IRIS 313-9482	H1	54	IRIS 313-10000	H1	84	IRIS 313-10041	H1	114	IRIS 313-10171	H1	144	IRIS 313-11171	H2
25	IRIS 313-9609	H1	55	IRIS 313-7832	H2	85	IRIS 313-11817	H1	115	IRIS 313-9065	H1	145	IRIS 313-11478	H2
26	IRIS 313-9758	H1	56	IRIS 313-10177	H1	86	IRIS 313-10863	H1	116	IRIS 313-8697	H1	146	IRIS 313-11166	H2
27	IRIS 313-11870	H1	57	IRIS 313-9594	H1	87	IRIS 313-12013	H1	117	IRIS 313-11361	H1	147	IRIS 313-9204	H3
28	IRIS 313-8994	H1	58	IRIS 313-9605	H1	88	IRIS 313-9882	H1	118	IRIS 313-10962	H1	148	IRIS 313-10593	H2
29	IRIS 313-11568	H1	59	IRIS 313-9433	H1	89	IRIS 313-8660	H1	119	IRIS 313-9148	H1	149	IRIS 313-11065	H2
30	IRIS 313-11052	H1	60	IRIS 313-8985	H1	90	IRIS 313-9560	H1	120	IRIS 313-9484	H1	150	IRIS 313-10682	H3

Table 2. Descriptive statistics for the studied population

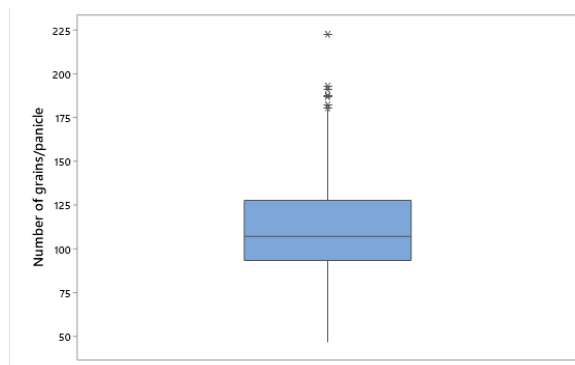| Statistics	Whole population
Range	47-223
Mean	113.34
Median	107.33
SE mean	2.53
CV(%)	27.35

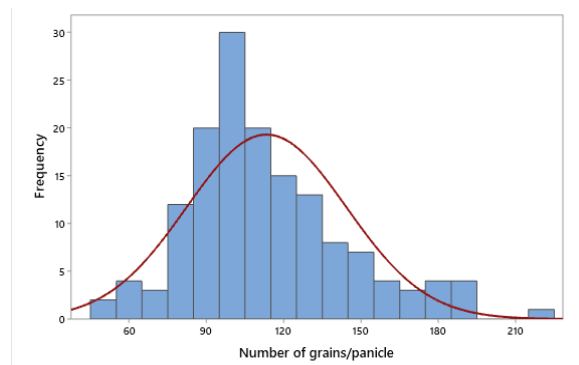**Fig.1. Box plot curve for number of grains**

Fig. 2. Frequency distribution (Histogram) per panicle curve for number of grains per panicle

Table 3. Allelic diversity analysis for *An-1* gene in 150 rice genotypes

REGION	SNP POSITION	ALLELE	TYPE	MAF
EXON 1	16734759-61	3bp	INDEL1	4.00
EXON 1	16734725	G/A	SNP1	6.67
EXON 2	16734542	G/C	SNP2	10.67
EXON 2	16734484	C/G	SNP3	8.00
EXON 2	16734476	1bp/0bp	INDEL2	48.67
EXON 2	16734442	C/A	SNP4	1.33
INTRON 2	16734402	0bp/1bp	INDEL3	6.67
INTRON 2	16734378	1bp/0bp	INDEL4	0.67
INTRON 2	16734332	C/A	SNP5	0.67
INTRON 2	16734244	C/A	SNP6	0.67
EXON 3	16734226-237	12bp/0bp	INDEL5	0.67
EXON 3	16734204-218	15bp/0bp	INDEL6	0.67
INTRON 3	16733811	C/T	SNP7	5.33
INTRON 3	16733609	1bp/0bp	INDEL7	5.33
INTRON 3	16733608	1bp/0bp	INDEL8	6.00
INTRON 3	16733607	1bp/0bp	INDEL9	0.67
INTRON 3	16733493	C/T	SNP8	7.33
INTRON 3	16733434	A/T	SNP9	9.33
EXON 4	16733302	A/G	SNP10	2.00
INTRON 5	16732975	G/A	SNP11	0.67
EXON 6	16732882	C/T	SNP12	1.33
EXON 6	16732870	T/C	SNP13	7.33
EXON 6	16732841	G/A	SNP14	1.33
EXON 6	16732750	T/G	SNP15	6.67
EXON 6	16732748	G/A	SNP16	6.67
EXON 6	16732732	C/T	SNP17	7.33
EXON 6	16732701	G/A	SNP18	8.00
INTRON 6	16732590	1bp	INDEL10	0.67
INTRON 6	16732582	A/T	SNP 19	5.33
INTRON 6	16732581	T/A	SNP 20	9.33

MAF - minor allele frequency

Apart from these missense mutations, even a nonsense mutation occurred at 16734476 positions of the nucleotide sequence (G deletion) leading to the frameshift mutation and production of the premature stop codon, which was analysed with ExPASy translate tool (http://expasy.org/tools/dna.html). The truncated protein consists of 97 amino acids and loses its functions by losing its bHLH domain (**Fig. 3**). In the first exon, a 3bp insertion causes insertion of alanine in the encoded protein. Since the *An-1* gene is a transcription factor with transactivation activity, all these changes have an impact on the protein produced and finally the yield. Luo *et al*. (2013) findings state that a transposon-like indel in the promoter region of Nipponbare subhaplotype *an-1*(Tn+) and another sub-haplotype *an-1*(G-) with 1-bp nucleotide-G deletion in the second exon of *An-1* led to frameshift mutation that enhances grain number per panicle comparatively and hence in our study the genotypes with G deletion in the second exon can be exploited further along with the consideration of non-synonymous SNPs.

Haplotype analysis of the *An-1* gene (LOC_Os04g28280) divided all the genotypes (150) into four different haplotype groups (**Table 1, Fig. 4**). Vasumathy and Alagu, (2021) identified three haplotype clusters in the third hap-block for *OsLG3* (grain yield and length) and twelfth hap-block for *OsMFT1* in chromosome 3 and 6, respectively. Gouda *et al*. (2020) obtained five haplotype groups for the *Gn1a* gene while considering forty-eight genotypes. *OsSPL14* had 15 haplotypes whereas, *OsMADS1* had only one haplotype across 3K RG panels (Abbai *et al.,* 2019). Among the various SNPs of the *An-1* gene, only seven significant SNPs were used for constructing four haplotypes using haploview as shown in **Fig. 4**. The Linkage Disequilibrium block depicts the alleles at nearby positions, that can occur on the same haplotype often than expected by chance (**Fig. 5**). Linkage Disequilibrium refers to the non-random association of alleles in a loci antagonist to Mendel's law of independent assortment. All pairwise combinations of SNPs yield high LD *i.e.* the linked SNPs have high LD. The SNP sites 10416732870

MNPTTAAAADQPSKPSAAAAARKRKSSAKPKASSSSLPTATATTNASPKRSKVAAGA
GDDGDADADAAEEKPEPAKDYIHVRASGGKPPIAIASPRG–GGRG–ARG–
SFCSRSCQAATRSPARLSCWTRSSTMCSRCSVRSSFCP–SWRP–
ILSWTLTAITCLPKIAICQYPHTRQAIRPPPPRSPTPAHPPLLIHSPSTTAGSSTSTP
LCKWEPPPDSAKTVQSQRWHPLPRHCRTILLFTASTGGSSSRGRQ–TT–RPSH

Fig. 3. Insilico translation shows that 'G' base deletion in *An-1* gene at 16734476 position results in protein truncation/premature termination

Block 1

GCTTCGG
ATGCACA
GCTTCCG
ATGCCCA

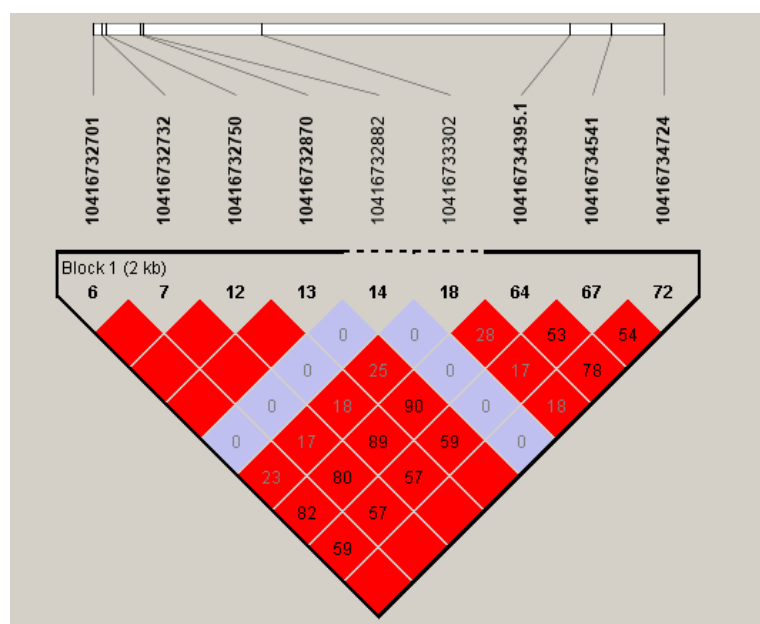**Fig. 4. Formation of haplotype groups based on allelic variation**

Fig. 5. Linkage Disequilibrium (LD) block of *An-1* gene for 150 genotypes. Values inside the block indicates LD r².

and 10416734395.1 had the highest LD r^2 value (90) indicating a high correlation between these two SNPs. From **Table 4**, the SNPs, their respective position and minor allele frequency could be easily identified. Among the four combinations, the H1 occupies the maximum number of genotypes (132) followed by H2 (10) and H3 (6). H4 occupies the least number of genotypes (2). It can be seen that the maximum number of grains per panicle falls under H1 (223) followed by H2 (124.67) and H3 (113) and the coefficient of variation is high for all the haplotypes except H4 (**Table 5**). Based on the median value from the descriptive statistics, H1 had 68 genotypes with more than 108 grains per panicle, whereas H2 and H3 had only two genotypes, which validates the superiority of H1

over other haplotypes. Further, Dunnett's test revealed that there exists a significant difference between H2-H1, H3-H1 (p value < 0.05) but there is no significant difference between H4-H1. Although theoretically, there should be a significant difference between H4-H1, due to the least number of genotypes under this haplotype, it's not significantly different (**Table 6 & Fig. 6**). In H1, the genotype SIGARDIS performed better with the highest number of grain numbers (223) and in H2 (124.67), GENIT yielded more grain numbers followed by DAMNOEUB KAUN KHMOM in H3 (113), thus resulting in increased yield. The lowest grain number was recorded in PA KHENG and BORO with 47 and 51 grain numbers per panicle, respectively. Hence, these genotypes can be

Table 4. Details of identified significant alleles.

Marker number	SNP site	Position	HW p-val	MAF	Alleles
6	**10416732701**	**16732701**	**1.9491E-18**	**0.080**	**G:A**
7	**10416732732**	**16732732**	**2.7365E-17**	**0.074**	**C:T**
12	**10416732750**	**16732750**	**3.8441E-16**	**0.068**	**T:G**
13	**10416732870**	**16732870**	**2.5338E-17**	**0.074**	**T:C**
14	10416732882	16732882	0.000069416	0.014	C:T
18	10416733302	16733302	1.1452E-06	0.020	A:G
64	**10416734395**	**16734395**	**3.1186E-16**	**0.067**	**C:A**
67	**10416734541**	**16734541**	**2.7123E-22**	**0.108**	**G:C**
72	**10416734724**	**16734724**	**3.5835E-16**	**0.068**	**G:A**

HW p-val - Hardy-Weinberg p value

Highlighted SNPs are responsible for haplotype formation

Table 5. Significant difference between haplotypes based on number of grains/panicle

Statistics	H1 (132)	H2 (10)	H3 (6)	H4 (2)
Range	75.67-223	51-124.67	47-113	72.3-94
Mean	117.3	84.4	84.4	83.2
Median	108.67	80.67	86.8	83.2
SE mean	2.58	8.37	11.4	10.8
CV (%)	25.3	31.36	33.05	18.42

H - haplotype

Table 6. Dunnett's Simultaneous Tests for Level Mean-Control Mean

Difference of levels	Mean difference	T-value	Adj. p-value
H2-H1	-32.9	-3.42	0.002
H3-H1	-32.9	-2.68	0.024
H4-H1	-34.1	-1.63	0.281
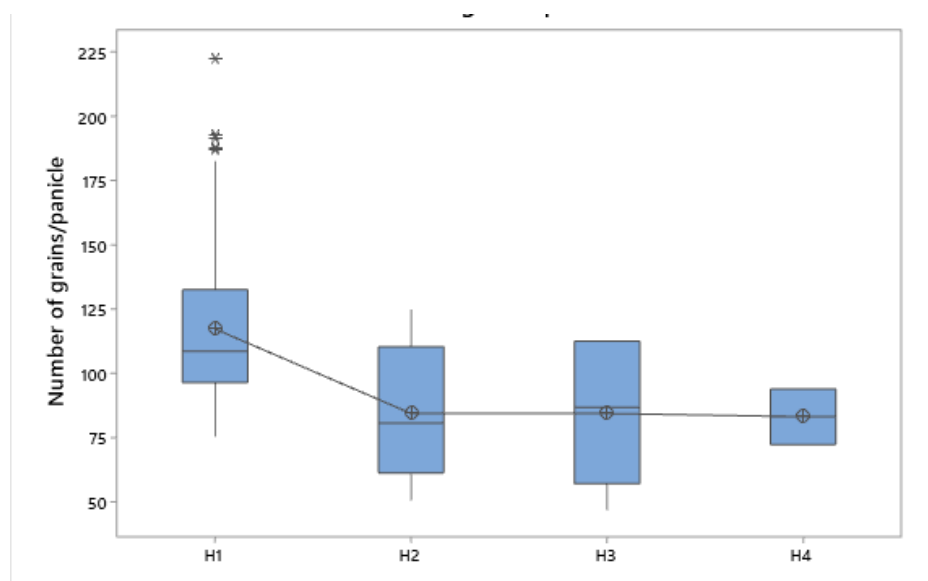
Fig. 6. Box plot curve for different haplotypes indicating significant difference by Dunnett's test.

recommended as suitable donors for developing mapping populations in the crop improvement programme. Further, analysis is required to identify the causative SNPs.

Overall, this study resulted in unravelling the allelic variations in the *An-1* gene with the help of 150 genotypes from 3K RG panel, and their grouping based on haplotyping identified a superior haplotype H1 and a further selection of genotypes from these haplotypes can be exploited in allele mining and plant breeding programmes.

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