Rice productivity has considerably improved due to the effective employment of heterosis, but the genetic basis of heterosis for grain shape and weight remains uncertain. For studying the genetic dissection of heterosis for grain shape/weight and their relationship with grain yield in rice, quantitative trait locus (QTL) mapping was performed on 1,061 recombinant inbred lines (RILs), which was developed by crossing xian/indica rice Quan9311B (Q9311B) and Wu-shan-si-miao (WSSM). Whereas, BC1F1 (a backcross F1) was developed by crossing RILs with Quan9311A (Q9311A) combined with phenotyping in Hefei (HF) and Nanning (NN) environments. Overall, 114 (main-effect, mQTL) and 359 (epistatic QTL, eQTL) were identified in all populations (RIL, BC1F1, and mid-parent heterosis, HMPs) for 1000-grain weight (TGW), grain yield per plant (GYP) and grain shape traits including grain length (GL), grain width (GW), and grain length to width ratio (GLWR). Differential QTL detection revealed that all additive loci in RILs population do not show heterotic effects, and few of them affect the performance of BC1F1. However, 25 mQTL not only contributed to BC1F1’s performance but also contributed to heterosis. A total of seven QTL regions was identified, which simultaneously affected multiple grain traits (grain yield, weight, shape) in the same environment, including five regions with opposite directions and two regions with same directions of favorable allele effects, indicating that partial genetic overlaps are existed between different grain traits. This study suggested different approaches for obtaining good grain quality with high yield by pyramiding or introgressing favorable alleles (FA) with the same direction of gene effect at the QTL regions affecting grain shape/weight and grain yield distributing on different chromosomes, or introgressing or pyramiding FA in the parents instead of fixing additive effects in hybrid as well as pyramiding the polymorphic overdominant/dominant loci between the parents and eliminating underdominant loci from the parents.
Introduction

Rice (*Oryza sativa*) is an important staple food globally, feeding about half of the world’s population. To fulfill the increasing feeding demand of the people, more attention should be paid to developing high-yield rice varieties (Ikeda et al., 2013). However, with the development of the economy and living standards, grain quality has also become the prime objective of breeding, along with high yield (Zou et al., 2018; de Oliveira et al., 2020). Grain weight is commonly evaluated by 1000-grain weight (TGW) that contributes to grain shape and yield traits, respectively (Qiu et al., 2017; Xia et al., 2017). Rice yield and grain quality are also determined by grain shape and weight (Yu et al., 2018). Thus, they significantly play an important role in rice breeding program because of their contribution to rice yield and quality.

The complex genetic mechanism of multiple grain traits is regulated by QTL, which locates on all 12 chromosomes (http://www.gramene.org/). Over the last decades, QTL mapping has been widely used to analyze grain shape quality traits (Bai et al., 2010; Han and Huang, 2013; Huang et al., 2013). By using different mapping populations (F₂, BC₁F₁, RIL, double haploid, DH), many QTL related to grain traits were identified (Huang et al., 2013; Kabange et al., 2020; Li et al., 2021). A few genes were cloned among them like GW2 (Song et al., 2007), GW5 (Liu et al., 2017), GW6 (Shi et al., 2020), GW8 (Wang et al., 2012), GS3 (Mao et al., 2010), GSS (Li et al., 2011), GS2 (Che et al., 2015; Hu et al., 2015), GS9 (Zhao et al., 2018), GLW7 (Si et al., 2016), GL4 (Wu et al., 2017), GL6 (Wang et al., 2019a), and LGY3 (Liu et al., 2018), which regulates grain weight and shape. Combining favorable alleles (FA) of these genes can enhance rice quality and yield. For instance, introgression of different alleles of LGY3 and GS3 into the indica variety RD23 can attain different yield-increasing effects, and pyramiding of LGY3 and GS3 in RD23 proved to be 10.9% more productive with improved grain quality than pyramiding of LGY3 and GS3 (Liu et al., 2018). Currently, single nucleotide polymorphisms (SNPs) are preferably used because of their abundant presence (Kumar et al., 2012), while next-generation sequencing (NGS) permits the detection of several SNPs for different plant species. Based on SNP marker, a high-density map can be prepared, which can improve the accuracy/efficacy of QTL mapping (Li et al., 2014).

Heterosis, a phenomenon, in which F₁-hybrids show far superior phenotypic performance compared to their parents (Chen et al., 2018). F₁-hybrid plants have noticeably increased grain yield, biomass production, vegetative growth rate, and stress tolerance (Fujimoto et al., 2018). Many hypotheses related to heterosis have been proposed as heterosis plays a crucial part in crop genetic research, such as dominance, over-dominance, and epistatic hypothesis (Hochholdinger and Hoecker, 2007; Goff and Zhang, 2013). Despite being broadly exploited in plant breeding, the molecular mechanism of heterosis is still poorly understood (Bagheri and Jelodar, 2010; He et al., 2010; Kim et al., 2017). Generally, heterosis is regulated by several genes, epistasis, and gene-by-environment interaction (Li et al., 2015; Li et al., 2018; Lin et al., 2020a). Researchers have extensively studied the heterosis in rice (Zhou et al., 2012; Xiang et al., 2016), but less attention has been paid to multiple rice grain traits.

A high-yielding superior Quan-you-si-miao (QYSM) hybrid derived from Q9311A and WSSM has good quality and yield, therefore it is essential to elucidate the genetic mechanisms/relationship of heterosis for multiple rice grain traits. Herein, our objective was to dissect the genetic basis of heterosis for grain weight/shape, and their heterotic relationships with grain yield using the same sets of the RIL and BC₁F₁ populations. We detected 126 QTL for multiple grain traits (grain type, weight, and yield) along with previously reported genes in these QTL. Besides, the heterotic effect of QTL for multiple grain traits (shape, weight, and yield) in RILs, BC₁F₁s, and H₃S₃S₃s was investigated to study the relative significance of both additive/non-additive effects on heterosis. Thus, our findings provide guidance about the genetic basis of heterosis for multiple grain traits that offers supportive information for effectively manipulating the identified QTL to improve the rice grain quality and yield.

Materials and methods

Experimental materials

QTL mapping was performed using 1,061 F₈ RILs derived from the cross of Quan9311B (Q9311B) and Wu-shan-si-miao...
Field experiment and trait evaluation

The RIL and BC1F1 populations were produced in the experimental field of Chinese Academy of Agricultural Sciences, Sanya (18.3°N, 109.3°E), Hainan province, in the 2018 winter season. The RILs, BC1F1s, parents (Q9311B and WSSM), and QYSM were sown in the seedling nursery on May 17, and July 12, 2019 in Hefei (32.3°N, 117.6°E) of Anhui province and Nanning (22.1°N, 107.5°E) of Guangxi province, respectively. Owing to its infertility, Q9311A was not included in the field experiment for traits evaluation, and its values were evaluated using the near-isogenic line Q9311B. The 25-day-old seedlings were transplanted into a four-row plot with six plants in each row at 13 × 30 cm spacing for two replications in a randomized complete block design. Field, water, weed, and fertilizer management followed to local practice.

At the mature stage, seeds from six uniform plants in the middle of each plot were harvested at the mature stage and air-dried for three months in drying houses to a 13.5% grain moisture content. Post eradicating the unfilled grains, the grains were weighed to calculate grain yield per plant (GY, g). Grain shape and grain weight traits: grain length (GL, mm), grain width (GW, mm), grain length-width ratio (GLWR), and 1000-grain weight (TGW), were measured with a scanning machine (SC-E, Wanshen Technology Company, Hangzhou, China). All measurements were performed with two replicate samples, and the mean was used for data analyses.

SNP genotyping

Genomic DNA for SNP genotyping was isolated from approximately 100-mg fresh leaf samples of 5-week-old seedlings for each of the 1061 RILs, Q9311B and WSSM using a modified cetyltrimethylammonium bromide method (Murray and Thompson, 1980). Whole-genome shotgun sequencing of the parental lines (Q9311B and WSSM) and RILs generated 150 bp paired-end reads consisting of ~2.4 Tb of sequence. The RILs were sequenced with ~10x genome coverage, and the two parental lines were sequenced with ~40x genome coverage using the Illumina sequencing platform Novaseq 6000 (CapitalBio Technology Inc., Beijing, China). The sequence reads of the parental lines, and RILs were aligned against the Nipponbare reference genome (IRGSP 1.0) (Kawahara et al., 2013) with BWA package 0.7.1 using default parameters, and PCR duplicates were removed by the ‘MarkDuplicates’ module in the Picard tools (version 1.119) (Li and Durbin, 2009). The raw reads were also re-aligned to identify highly polymorphic regions using the ‘IndelRealigner’ function in GenomeAnalysisTK 3.4.0 (DePristo et al., 2011). Sequence variants between the parental lines were called with ‘UnifiedGenotyper’ in GenomeAnalysisTK. Only uniquely mapped reads were used for subsequent SNP calling. Genotype calling of each RIL was performed based on the SNP alleles between the parents. A total of 741,928 homozygous SNPs polymorphic between Q9311B and WSSM were used for SNP calling in the 1061 RILs. The SNPs were further filtered by removing minor allele frequencies < 0.01 or missing rates > 0.2, leaving 156,373 high-quality and evenly distributed SNPs, located within the genic regions of 15,043 genes based on the gene annotation of the Nipponbare RefSeq from RAP-DB (released on June 26, 2019) (Sakai et al., 2013). A set of 1196 genes was removed because one of the two parental alleles was rare or showed a high frequency of the heterozygote in the RIL population. Then, an allelic genotyping dataset containing two genotypes (Q9311B type and WSSM type) at each of 13,847 genes was generated to be used for bin map construction for the 1061 RILs.

Bin map construction and QTL mapping

A bin map is a type of genetic map constructed using bin (binning of redundant markers) genotypes, which can save running time of linkage and QTL mapping by greatly reducing numbers of redundant markers. The bin map of the RIL population was constructed based on the two parental genotypes at 13,847 genes using the BIN function in IciMapping QTL version 4.2 (Meng et al., 2015). To remove redundancy, only a single gene was retained to represent each bin, either one gene with a minimum missing rate or random gene when the missing rates were equal. Genes with unique genotypes (those that did not belong to a redundant bin) were excluded from bin-map construction. As a result, 855 bins were identified with mean size ~440 kb and used for construction of bin map using the BIN function in IciMapping QTL version 4.2 (Meng et al., 2015). Allelic genotyping datasets containing two bin genotypes (Q9311B type and WSSM type) at each of 855 bins for the 1061 RILs, and the genotype for each BC1F1 hybrid were deduced from the gene-based allelic genotypes of its parental RIL and Q9311B. Specifically, for each of the 13,847 genes, if the parents (RIL and Q9311B) have the same
allelic genotype, their BC1F1 hybrid should be the homozygous genotype of Q9311B; if the parents had different homozygous genotypes, the allelic genotype of their BC1F1 hybrid was deduced as the heterozygote; and if RIL parent had the heterozygous genotype, the BC1F1 was treated as missing. The bin genotypes were used for QTL mapping.

Considering that segregation of restoring gene in RIL population can result in a difference in fertility in BC1F1 population derived from Q9311A x RILs, we speculated a RIL had restoring gene if its test-cross (BC1F1) with Q9311A had more than 70% fertility. Based on this criterion, 793 and 797 RILs were selected as restorer lines in HF and NN, respectively, because of their corresponding BC1F1 having fertility of more than 70%. So, QTL mapping for four traits (three grain shape traits and TGW) was performed separately for 1061 RIL and the corresponding BC1F1 and HMP populations evaluated in both environments, while QTL mapping of GYP was separately conducted using 793 and 797 RILs with restoring ability and their corresponding BC1F1 and HMP populations in HF and NN, respectively. The HMP population was produced by estimation of mid-parental heterosis for each BC1F1 as follows, the mid-parental heterosis value, HMP (%) = (F1−MP)/MP×100, where F1 is the trait value of BC1F1, and MP is the mean value of the corresponding paternal RIL and Q9311B (Lariépe et al., 2012). All traits were analyzed using the BIP (biparental populations) function in IciMapping QTL. The population settings of RILs and BC1F1s were F1 RIL and P2BC1F1, respectively. The inclusive composite interval mapping of additive (ICIM-ADD) QTL method was performed to identify main-effect QTL (mQTL) using the default settings in which a QTL were retrieved based on the left and right markers of the detected interval.

Inference of mQTL gene actions

The gene effects of heterosis can be dissected through QTL mapping using segregation populations such as the RILs derived from QYSM and the BC1F1s between Q9311A and the RILs, where the additive effect (a) of mQTL can be detected in the RILs while the sum of additive effect and dominance effect (a+d) detected in BC1F1s. When HMPs are used for QTL analysis of heterosis, the inherited effect should be a dominance effect (d). The identified mQTL can be divided into four types in RILs and BC1F1s according to Mei et al. (2005) and Kim et al. (2017). Thus, when a QTL was observed only in the RILs or BC1F1s, it was called additive effect (A). When QTL were simultaneously detected in the RILs and BC1F1s, or in the HMP of BC1F1s, QTL with |2a/(a + d)|<1 or |2a/(a + d)|>1 were designated as incomplete dominant QTL (ID). QTL with |2a/(a + d)|>1 or |2a/(a + d)|<1, or those detected only in HMP datasets were designated as overdominant QTL (OD), and QTL showed negative values were defined as underdominant QTL (UD). Only ID, OD, and UD mQTL were used for subsequent heterosis analysis of the detected mQTL.

Identification of candidate genes

QTL with heterotic effects simultaneously detected in both environments for grain shape, grain weight, and grain yield traits were used for candidate gene analysis. All genes in each QTL interval were identified for further candidate-gene analysis using haplotype analysis of 732 accessions (386 indica/xian, 219 japonica/geng, 46 of intermediate type, 67 aus/boro, and 14 basmati/sadri accessions) from the 3K Rice Genome Project (3K RGP) (Supplementary Table 1). Grain shape, TGW and GYP phenotypic data were collected in Sanya. All available high-quality SNPs with a minor-allele frequency of more than 0.05 and a missing rate of less than 20% located in these genes were retrieved from 18 M SNP data generated from 3K RGP in the Rice SNP-Seek Database (Wang et al., 2018). Haplotype analysis was conducted for each candidate gene in each QTL region using all non-synonymous SNPs located inside the gene CDS region. Haplotypes shared by more than 10 accessions were compared. For a single candidate gene, the haplotype with the best phenotype of the corresponding trait was assigned as a favorable haplotype.

Statistical analysis

Differences in the mean phenotypic values among the haplotypes were evaluated by one-way ANOVA by the agricolae package in R (de Mendiburu, 2021). Phenotypic differences among the check parents and the relative hybrids were assessed by Duncan’s multiple comparison test, and among RILs and the BC1F1s by Student’s t-test. HMP was tested with a Student’s t-test based on the contrast between the F1 hybrid mean and the mean performance of the corresponding parental lines (Lariépe et al., 2012). Phenotypic correlation analyses of the five traits were computed using the corplot package in R. Broad-sense heritabilities ($H^2$) were calculated using the AOV module implemented in IciMapping QTL.
Results

Phenotypic performances

Table 1 showed the field performance and broad-sense heritability of parental population (Q9311B and WSSM), QYSM (F1), the RILs, the BC1F1s (Q9311A×RILs), and the corresponding HMPs in HF and NN for grain shape/yield traits. Two parents, Q9311B and WSSM, are slender grain rice varieties with similar grain weight and yield except that WSSM showed significantly lower GW and GL but larger GLWR than Q9311B in HF and NN, respectively. The HMPs of QYSM was 4.45% and 4.48% for GL, 2.25% and 3.44% for GW, 1.72% and 0.46% for GLWR, 0.91% and 16.73% for TGW, and 8.69% and 3.00% for GYP in HF and NN, respectively, showing significant heterosis in TGW in NN. Both RIL and BC1F1 populations displayed wide segregations in five traits, particularly TGW and GYP in both environments. Comparing to QYSM, we found lower heterosis in BC1F1 population for GL, GLWR, and GYP in the two environments and GW in HF except GW in NN and TGW in the two environments, demonstrating hybrid breakdown in the RILs in GL, GLWR, and GYP because of inbreeding depression, as RILs of QYSM showed GL (96.46 and 96.55%), GLWR (99.35 and 98.73%), and GYP (88.26 and 96.25%) in HF and NN, respectively. The broad-sense heritabilities (H2) of RILs (BC1F1s) were 0.99 and 0.99 for each of GL, GW, and GLWR, 0.55 and 0.68 for TGW, and 0.24 and 0.48 for GYP in HF and NN, respectively (Table 1). Thus, the environment strongly affected the grain traits, particularly TGW and GYP.

The significant association among different traits (grain shape, weight, and yield yield) of RILs and BC1F1s under the two locations was identified by correlation analysis (Table 2). Under both environments, we found significant (positive and negative) correlations between GLWR and GL, and GLWR and GW in RILs and BC1F1s. Whereas, GL showed significant negative correlation with GW in BC1F1s but not in RILs. Similarly, TGW showed significant positive correlations with GL and GW in RILs but not in BC1F1s under NN, demonstrating that few correlations were related to genetic background. However, other correlations were linked to the environment. For example, TGW showed positive correlations with GL and GW in BC1F1s only in HF but not in NN, and it was true for a negative correlation between GYP and GL only in HF in BC1F1s. Whereas, all traits (GL, GW, and GLWR) presented significantly positive correlations in both RILs and BC1F1s between the two environments, but TGW only in RILs and GYP only in BC1F1s, indicating that GL, GW, and GLWR exhibited that their performance did not affect by the background and the environment.

Main-effect QTL (mQTL) for grain shape, weight and yield traits

We constructed 855-bin linkage map for RIL population of about 2,268 cM, with chromosome length ranging from 106 cM (chromosome 7) to 334 cM (chromosome 2) (Supplementary Table 2). About 114 mQTL were identified for different grain traits (shape, weight, and yield) of RILs, BC1F1s, and corresponding HMPs in NN and HF by employing genotypic datasets at 855 bins for the 1,061 RILs, BC1F1s and HMPs. Among 114, we found 39 in both locations (Figure 1; Supplementary Table 3).

### Table 1 Phenotypic performance and broad-sense heritability of grain shape and grain weight as well as grain yield traits in the parent lines, QYSM (Q9311A × WSSM), RILs, BC1F1s (Q9311A × RILs), and their HMPs in Hefei (HF) and Nanning (NN).

| Traita | Env. | Q9311Bb | WSSM | QYSM | RILs | BC1F1sb |
|--------|------|---------|------|------|------|---------|
|        |      |         |      |      | Mean ± SD | CV (%) |       | Mean ± SD | CV (%) |       | Mean ± SD | CV (%) |       |
|        | HF   | 6.31b   | 6.13c | 6.50a | 4.45 | 6.27 ± 0.32 | 5.12  | 0.99  | 6.44 ± 0.24 | 3.69  | 0.99  | 2.42 ± 3.43 |
|        | NN   | 6.45b   | 6.29c | 6.66a | 4.48 | 6.43 ± 0.33 | 5.12  | 0.99  | 6.56 ± 0.24 | 3.69  | 1.98 ± 4.35 |
|        | HF   | 2.25a   | 1.91c | 2.13b | 2.25 | 2.07 ± 0.10 | 4.91  | 0.99  | 2.21 ± 0.06 | 2.91  | 0.99  | 2.20 ± 3.13 |
|        | NN   | 2.10a   | 1.88b | 2.12a | 3.44 | 2.09 ± 0.10 | 4.56  | 0.99  | 2.21 ± 0.06 | 2.50  | 0.55 ± 2.96 |
|        | HF   | 2.81a   | 3.22b | 3.07c | 1.72 | 3.05 ± 0.21 | 7.05  | 0.99  | 2.93 ± 0.15 | 5.13  | 0.99  | 0.16 ± 4.96 |
|        | NN   | 3.10b   | 3.35c | 3.14a | 0.46 | 3.10 ± 0.22 | 7.07  | 0.99  | 2.98 ± 0.15 | 4.97  | -3.66 ± 4.66 |
|        | HF   | 30.88a  | 27.56b | 29.49c | 0.94 | 25.37 ± 2.58 | 10.17 | 0.55  | 27.71 ± 2.03 | 7.32  | 0.68  | 2.90 ± 7.12 |
|        | NN   | 27.60b  | 23.97b | 30.10a | 16.73 | 28.13 ± 2.94 | 10.45 | 0.55  | 32.17 ± 4.48 | 13.93 | 17.55 ± 16.81 |
|        | HF   | 26.03a  | 26.03a | 28.29c | 8.69 | 24.97 ± 3.90 | 15.61 | 0.24  | 24.70 ± 6.06 | 24.54 | 0.48  | -3.02 ± 6.30 |
|        | NN   | 41.32a  | 44.05a | 43.97c | 3.00 | 42.32 ± 8.01 | 18.92 | 0.48  | 43.18 ± 10.77 | 24.94 | -1.44 ± 11.92 |

aGL, grain length; GW, grain width; GLWR, grain length-width ratio; TGW, 1000-grain weight; GYP, grain yield per plant.
bDifferent letters represent significant differences at P < 0.05 based on Duncan’s multiple range test for the same traits among the two parents and QYSM.
c*represents significant difference at P < 0.05 based on t-test for the same traits between F1 and the mean of the two parents.
dH² represents broad-sense heritability of traits.
For four grain type and weight traits, total (21, 20) mQTL were detected in HF and NN for GL, explaining (34.0, 25.0)%,(35.6, 33.3)%, and (18.4, 16.3)% of total phenotypic variances in the RILs, BC 1F1s, and H MPs( Supplementary Table 3 ). These 41 mQTL included additive (25), over-dominance type (OD-t, 9), under-dominance type (UD-t, 3), and incomplete-dominance type (ID-t, 4) ( Figure 2 ). Among them, 15 mQTL were identified in both environments, such as three (qGL1.4, qGL7.1, and qGL7.2) in RILs, two (qGL9.1 and qGL9.3) in BC1F1s, and two (qGL11.1, qGL6.1) in H MPs with the same gene actions. While, eight QTL (qGL1.2, qGL2, qGL3.1, qGL5.1, qGL6.2, qGL9.4, qGL11.1, and qGL11.2) were simultaneously detected in two or three populations in different environments with same or different gene actions. For GW, 9 (HF) and 17 (NN) mQTL were identified, explaining (0 HF, 10.1 NN)%, (31.8 HF, 42.8 NN)%, and (12.3 HF, 8.2 NN)% of the total phenotypic variance in the RILs, BC1F1s, and H MPs (Supplementary Table 3 ). These 26 mQTL contained 16 additives, 1 ID-t, 4 UD-t, and 5 OD-t (Figure 2). Among them, 5 mQTL were concurrently seen in both HF and NN, including qGW2.3 in H MPs, qGW4.2, and qGW5 in BC1F1s, while qGW6.1 and qGW9.2 were found in BC1F1s (HF), and BC1F1s and H MPs in NN with different gene actions. A total of 16 (HF) and 23 (NN) mQTL for GLWR were identified, accounting for (13.9 HF, 25.5 NN)%, (38.0 HF, 42.8 NN)%, and (12.3 HF, 8.2 NN)% of the total phenotypic variance in the RILs, BC1F1s, and corresponding H MPs, respectively (Supplementary Table 3 ). These 39 mQTL comprised 26 additive, 2 ID-t, 6 UD-t, and 5 OD-t (Figure 2). Thirteen mQTL were commonly present in both environments, from which 7 mQTL, including (qGLWR2.1, qGLWR3.1, qGLWR3.2, qGLWR4.2, qGLWR9.1, and qGLWR9.2) were identified in the BC1F1s, and one mQTL (qGLWR8.2) in the RILs of HF and NN.

### TABLE 2 Correlations between grain shape and grain weight as well as grain yield traits estimated in RIL population and BC1F1 (Q9311A × RILs) population in Hefei (upper triangular) and Nanning (lower triangular).

| Trait | GL     | GW     | GLWR   | TGW    | GYP    |
|-------|--------|--------|--------|--------|--------|
| RIL population |        |        |        |        |        |
| GL    | 0.80** | -0.01  | 0.72** | 0.65** | 0.08   |
| GW    | -0.12  | 0.78** | -0.69**| 0.47** | 0.05   |
| GLWR  | 0.78** | -0.71**| 0.82** | 0.14   | 0.02   |
| TGW   | 0.29** | 0.40** | -0.04  | 0.43** | 0.07   |
| GYP   | 0.12   | 0.07   | 0.03   | 0.07   | 0.01   |
| BC1F1 population |        |        |        |        |        |
| GL    | 0.68** | -0.25* | 0.86** | 0.31*  | -0.20* |
| GW    | -0.26* | 0.61** | -0.71* | 0.22*  | -0.09  |
| GLWR  | 0.88** | -0.69**| 0.69*  | 0.14   | -0.09  |
| TGW   | 0.18   | 0.14   | 0.12   | 0.07   | -0.18  |
| GYP   | -0.16  | 0.01   | -0.16  | -0.02  | 0.38** |

The values on principal diagonal indicated correlations between Hefei and Nanning. The values were correlation coefficients (r). The * and ** indicated significant correlation at P < 0.05 and P < 0.01, respectively. GL, grain length; GW, grain width; GLWR, grain length-width ratio; TGW, 1000-grain weight; GYP, grain yield per plant.
with same gene action. However, the other 5 mQTL (qGLWR4.1, qGLWR5.2, qGLWR6.2, qGLWR6.4, and qGLWR9.3) were observed in two or three populations in both locations (HF and NN) with different gene actions. About TGW, 11 (HF) and 9 (NN) mQTL were identified, explaining (37.3 HF, 4.1 NN)%, (20.5 HF, 25.6 NN)%, and (7.9 HF, 16.1 NN)% of total phenotypic variance in the RILs, BC1F1s, and HMPs (Supplementary Table 3). These 20 mQTL contained 11 additive, 1 ID-t, 2 UD-t, and 6 OD-t (Figure 2). No any common QTL for TGW was detected in the two environments.

For GYP, 13 (HF) and 14 (NN) mQTL were detected, explaining (0 HF, 8.7 NN)% , (45.8 HF, 27.7 NN)% , and (38.8 HF, 37.9 NN)% of total phenotypic variance in the RILs, BC1F1s, and the corresponding HMPs (Supplementary Table 3). These 27 mQTL contained 9 additive, 2 UD-t, and 16 OD-t (Figure 2). Among these, 6 mQTL were simultaneously detected in both HF and NN locations, including qGYP4.1 and qGYP4.2 detected in HMPs and BC1F1s, respectively, qGYP3.2 and qGYP10.3 in both locations of BC1F1s and HMPs. However, qGYP1.2 was seen in BC1F1s and HMPs in HF, and HMPs in NN, while qGYP2.3 seen in BC1F1s in HF, and RILs and BC1F1s in NN with differing gene actions.

Furthermore, 27 mQTL (additive effects) were found in the RILs, 97 (additive and dominant effects) were found in BC1F1s, and 54 (dominant effects) were observed in HMPs (Supplementary Table 3). However, there was only one mQTL (qGL9.4 in HF) found between the RILs and HMPs, and among 11 mQTL, 4 (qGL2, qGL4.1, qGLWR4.1, and TGW9) in HF, 6 (qGL6.2, qGL11.1, qGW8, qGLWR9.3, qGYP1.1, and qGYP2.3) in NN, and one mQTL (qGLWR2.2) present both in HF and NN between RILs and BC1F1s, indicating that most of additive loci in the RILs did not exhibit heterotic effects as they did not simultaneously find in HMPs. The fixed favorable allele (FA) from WSSM or Q9311 from these 11 mQTL affect BC1F1s performance but did not show a significant heterotic effect.

Twenty-five mQTL were commonly observed between BC1F1s and HMPs, which recommended that these loci affect the performance of BC1F1s and also contributed to heterosis.

Epistatic QTL for grain shape, weight and yield traits

In total, 73, 164, and 122 eQTL pairs were observed in RILs, BC1F1s, and HMPs (Supplementary Tables 4, 5). Among these, 3, 36, and 11 pairs involved one mQTL and the others a random locus in the RILs, BC1F1s, and HMPs, respectively. In addition, 2 pairs involved two mQTL were found in BC1F1s in NN. The total phenotypic variation explained (PVE) of eQTL was 21.6%, 65.7%, and 27.3% (GL), 23.3%, 55.2%, and 18.3% (GW), 24.4%, 61.7%, and 32.8% (GLWR), 18.4%, 30.1%, and 25.1% (TGW), and 14.9%, 18.2%, and 6.1% (GYP) in HF in the RILs, BC1F1s, and HMPs. In NN, 28.9%, 53.8%, and 29.6% (GL), 24.7%, 49.6%, and 13.1% (GW), 25.3%, 56.9%, and 30.2% (GLWR), 10.9%, 24.4%, and 24.4% (TGW), and 15.3%, 1.6%, and 0% (GYP) were found in RILs, BC1F1s, and HMPs, implying that epistasis is a key factor in heterosis. About 23.64% of common epistasis was detected for TGW between BC1F1s and HMPs in the two locations, which is significantly higher than 4.35% (GL), 0% (GW), 3.80% (GLWR), and 0% (GYP), demonstrating that epistasis displayed higher contribution in grain weight with heterotic effect as compared to grain type and yield traits in BC1F1s.

QTL regions simultaneously affecting grain shape, weight and yield traits

We found seven QTL regions on chromosomes (chr) 1, 2, 3, 4, 5, 10, 11 that simultaneously affected the grain shape, grain weight.
weight, and grain yield traits under the same environment (Figure 3; Supplementary Table 6). Five regions were found with opposite directions of FA effects of grain shape, weight, and grain yield traits, including the region of 6463296−6583687 bp on chr1, containing qGLWR1.2 and qGYP1.2 detected in BC1F1s and HMPs in HF; the region of 28248757−30003190 bp on chr3, harboring qGYP3.2 identified in BC1F1s and HMPs, while qGW3.6 and qGLWR3.4 detected in HMPs in NN; the region of 5111957−6692156 bp on chr4 harboring qGLWR4.2 and qGYP4.2 detected in BC1F1s in NN; the region of 4578959−5986015 bp on chr5 harboring qGW5, qTGW5 and qGYP5 detected in BC1F1s in NN; and the region of 20001168−20792875 bp on chr10 harboring qGL10 and qGYP10.4 detected in BC1F1s in NN. However, we found two regions with the consistent direction of FA effects for grain traits (type and yield), including 8135122−8389827 bp on chr2, containing qGLWR2.1 and qGYP2.2 found in BC1F1s in HF, and the region of 8243782−9022284 bp on chr11 harboring qGL11.2 detected in HMPs, while qGYP11 detected in BC1F1s and HMPs in NN. Based on this, we assume that a partial genetic overlap is existed between different grain traits (shape, weight, and yield) either in same or opposite directions of FA for each mQTL.

Candidate genes affecting grain shape, weight and grain yield

Among 39 mQTL affecting rice grain (shape, yield) traits in both environments (Supplementary Table 7), 12 mQTL had heterotic effects without previous cloned genes. Based on haplotype analysis, there was no significant difference in haplotypes among qGLWR2.2, qGYP4.1, and qGL11.1. In the remaining 9 mQTL, 19 candidate genes with haplotype differences were detected. While, two different populations (xian, geng) were used for phenotypic characterizations of the haplotypes (Supplementary Table 8).

We found four candidate genes (Os01g0218032, Os01g0217050, Os01g0217433, and, Os01g0217400) for qGYP1.2 (Figure 4; Supplementary Table 8). Five and three haplotypes were found in both (xian, geng) populations for Os01g0217050. Besides, Hap 1 showed profound increase in GYP but decrease GL, GW, GLWR and TGW only in geng population. Similarly, five and two haplotypes were found in both populations for Os01g0217400. While Hap 1 showed significant increase in GYP but decrease GL, GLWR and TGW only in geng population. In xian population, both genes Os01g0217050 and Os01g0217400 had prominent effects on GYP, GL, GW and GLWR but did not show any effect on TGW. Four and two haplotypes were found in both populations for Os01g0217433, and Hap 3 markedly increased GYP, GL, and GLWR only in xian population. In contrast, Hap 1 markedly increased GYP but decreased GL, GW, and TGW only in geng population. Three haplotypes were found in both populations for Os01g0218032, and Hap 1 enhanced GYP but decreased GL, GW, and TGW only in geng population. Further, Os01g0218032 did not show any noticeable effect on grain shape/weight traits in xian population.

![Figure 3](https://example.com/fig3.png)

**Figure 3**: mQTL affecting grain shape and grain weight as well as grain yield traits detected in BC1F1s and the corresponding HMPs in Hefei (HF) and Nanning (NN). Number on the left side of the chromosome indicates the genetic position of the QTL in cM. The linkage map spans 2,268 cM with a mean of 2.65 cM between neighboring bins.
Four candidate genes (Os02g0670000, Os02g0670400, Os02g0671100, and Os02g0671800) were observed for qGYP2.3 (Supplementary Table 8). Two haplotypes for Os02g0670000, Os02g0670400, and Os02g0671100 were found in both populations. The favorable Hap 1 in two populations markedly enhanced GYP but did not exhibit any effect on TGW in xian population, while showed substantial effect on GW in geng population. In xian population, five haplotypes were observed for Os02g0671800 and the Hap 4 had considerably enhanced GYP, GL, and GLWR. No haplotype for Os02g0671800 was detected in the geng population.

For qGYP3.2, four candidate genes (Os03g0715600, Os03g0716400, Os03g0717600, and Os03g0719500) were identified only in xian population (Supplementary Table 8), suggesting this gene is specific to xian subspecies. Four (three) haplotypes were observed for Os03g0715600 (Os03g0716400), and Hap 2 considerably enhanced GYP, GL, and GLWR but decreased GW and TGW in xian population. Both
Os03g0716400 and Os03g0719500 contain two and four haplotypes, and Hap 2 considerably enhanced GYP, GL, and GLWR but did not show any effect on TGW in xian population.

For qGLWR4.1, three candidate genes (Os04g0165400, Os04g0168100, and Os04g0168966) were observed (Figure 5; Supplementary Table 8). Five and two haplotypes were found in both populations for Os04g0165400, the favorable Hap 1 considerably enhanced GLWR, GL, TGW, and GYP but decreased GW in the xian population. In geng population, Os04g0165400 showed prominent effect on GL, TGW, and GYP. Four and three haplotypes were found in both populations for Os04g0168100, and Hap 3 had noticeable effect on GL, GW, GLWR, and GYP in the two populations. Five and two haplotypes were found in both populations for Os04g0168966, and Hap 1 had significant effect on GW, GYP, GLWR, and TGW only in xian population. Whereas, Os04g0168966 presented noticeable effect on GYP, TGW, and GL only in geng population.

Two candidate genes (Os05g0567500 and Os05g0569300) were detected for qGL5.1 and qGLWR5.2 (Supplementary Table 8). Whereas, we found only two haplotypes in geng population, suggesting this gene is specific to geng subspecies. Hap 1 had significant effect on GL, GW, and GLWR for Os05g0567500 and Os05g0569300 in the geng population.

For qGL6.1, qGW6.1, and qGLWR6.2, two candidate genes (Os06g0214900 and Os06g0215200) were identified (Supplementary Table 8). Three haplotypes for Os06g0214900 and two haplotypes for Os06g0215200 were observed in the two populations. A considerable effect was observed in GL, GW, GLWR, and GYP for Hap 3 of Os06g0214900 in both populations. For Os06g0215200, Hap 2 presented noticeable effect on GL, GLWR, and GYP in xian population, and GL, GW, GYP, and GLWR in geng population.

These findings indicated that many candidate genes have pleiotropic effects with different favorable haplotypes between these two populations, verifying that these genes are linked with variations in grain traits (shape, yield) between the two subspecies.

Discussion

The simultaneous improvement of cereal crops grain quality and yield is a main challenge in modern agriculture. Grain weight and grain type traits that contribute to grain yield greatly influence the commercial value of rice (Zheng et al., 2007; Qiu et al., 2015). Nowadays, breeding objectives of hybrid rice have been transferring from high yield to synergy of high yield and premium quality. So, it is vital to dissect genetic mechanisms of heterosis and their genetic relationships for effectively developing rice hybrid with good grain quality and high yield. QYSM is a best choice for genetic dissection of heterosis for rice grain traits.

Genetic dissection of heterosis for grain type, weight and grain yield

Despite the extensive reports on heterosis (Chu et al., 2012; Kim et al., 2017; Yang et al., 2021), reports about heterosis of grain traits are very rare. In this work, single-locus overdominance plays a critical role in rice yield heterosis (Figure 2), in accordance with the results reported by Melchinger et al. (2007) and Li et al. (2015). Contrarily, single-locus overdominance and epistasis affected rice grain shape heterosis (Figure 2; Supplementary Tables 4, 5). Five OD-t mQTL (qGL3.1, qGL11.2, qGW2.3, qGLWR6.3, and qGLWR11.2), and two ID-t mQTL (qGL2 and qGL6.2) were detected as heterotic loci for grain shape and mapped together with six known genes related to grain shape, i.e., SRL2 (Liu et al., 2016), SRS5 (Segami et al., 2012), SGL1 (Wang et al., 2016), TGW2 (Ruan et al., 2020), GL6 (Wang et al., 2019a), and GW6a (Song et al., 2015) (Supplementary Table 3), suggesting above cloned genes may have heterosis for grain type traits. The moderate rolled leaf is ideal for rice phenotype as it is helpful to attain exemplary plant architecture, consequently improving photosynthesis efficiency and grain yield (Wu, 2009). SRL2 gene plays a crucial role in regulating leaf development, and its deficiency leads to inwardly rolled leaves (Liu et al., 2016). Furthermore, grain type and grain weight are mainly correlated to rice yield potential. SRS5 identified with an overdominance effect on heterosis of GL, GLWR, and GYP promotes the rice grain size (Si et al., 2016; Segami et al., 2017), and positively regulates the GL through cell elongation (Segami et al., 2012). Similarly, SGL1 also increases the GL of rice (Wang et al., 2016). GW6a and GL6 regulate the grain weight, while TGW2 negatively regulates the grain weight (Ruan et al., 2020).

Similarly, four UD-t mQTL (qGL11.1, qGLWR11.1, qGL9.4 and qGLWR9.3) and three OD-t QTL (qGW9.2, qTGWR1.1, and qGYP10.3) were detected for heterosis of grain shape, weight, and grain yield and mapped together with the known genes related to grain yield like D2 (Dong et al., 2016), negative regulators of grain yield gene Gnl (Ashikari et al., 2005; Li et al., 2016), grain width, and weight GWS1L (Tian et al., 2019), and two other cloned genes related to grain type, plant height, and grain numbers, such as OsBCI (Jang et al., 2017), and Ghd8 (Wang et al., 2019b). Tiller angle is a vital part of plant architecture that significantly affects the grain yield. A tiller angle-related gene D2 positively improves rice plant architecture by controlling tiller angle and enhancing grain yield (Dong et al., 2016). However, GWS1L and Gnl negatively regulate the rice grain width, grain weight, and production, respectively (Li et al., 2016; Tian et al., 2019). In contrast, two genes, Ghd8 and OsBCI, positively regulate the grain number, grain size, and plant height (Jang et al., 2017; Wang et al., 2019b). Thus, these biological
FIGURE 5
Exon-intron structure and haplotype analysis of three candidate genes in whole, xian and geng populations for qGLWR4.1. (A) Os04g0165400, (B) Os04g0168100, (C) Os04g0168966. Different letters indicate significant differences in trait values among haplotypes by Duncan’s test at p < 0.05.
function genes influencing different grain traits may assess the genetic effect on heterosis of grain shape and yield.

Supplementary Table 5 indicated higher epistasis for grain weight than grain shape quality. Epistasis was compared between BC1F1s and HMPs in two environments, and more epistasis was found between BC1F1s and HMPs for grain weight than three grain shape and grain yield traits, indicating that epistasis had more contribution to heterosis of grain weight than grain shape and grain yield traits. By considering the number of mQTL identified for different grain traits, it is concluded that heterosis for TGW was mainly controlled by epistasis while GL, GW, GLWR, and GYP was mainly controlled by UD or OD, indicating that the genetic mechanisms is different for both grain shape and grain yield, and grain weight.

**Effect of environment on grain shape, weight and grain yield heterosis**

Crop yield heterosis is mainly affected by environment (Li et al., 2015; Lin et al., 2020a; Lin et al., 2020b). Previous study also revealed that the heterotic loci for yield were mainly affected due to genetic background and environmental conditions (Li et al., 2018). However, the behavior of environmental factors affecting rice’s heterosis is still uncertain. The current research showed a lower proportion (55%) of environment-specific mQTL for grain shape heterosis compared to yield heterosis (78%), while a high proportion (100%) of mQTL for grain weight heterosis was observed. Thus, the impact of the environment on heterotic loci of grain shape is considerably different from grain weight and yield. Such as qGL2, qGL3.1, qGL9.4, qGLWR4.1, qGLWR5.2, qGLWR6.2, and qGLWR6.4 were detected as the heterotic loci in HF but with additive effect in NN, while opposite results were observed in qGL5.1, qGL11.1, qGW6.1, qGW9.2, qGLWR9.3, and qGYP2.3 (Supplementary Table 3). Additionally, most eQTL were identified for heterosis of environment-specific grain shape traits (Supplementary Table 5). These results suggested that the environmental conditions affected heterotic loci of grain shape and weight in a different way.

**Comparisons of detected grain quality mQTL with previously known genes**

Rice grain quality affected by mQTL that was simultaneously found in both locations were compared with already reported genes present at the same or adjacent to the physical location. (Figure 1; Supplementary Table 3). Some identified mQTL were present near the already known genes. For example, qGL7.2 and qGLWR7 in the 22056455-26044575 bp region on chr 7, affecting GL in RILs under both environments, and GLWR in RILs in NN, were mapped together with OsGASR9 for grain size and yield in rice (Li et al., 2019), and GL7 for grain size (Wang et al., 2015). Similarly, qGW3.8 and qGLWR3.5 in the region of 34560598–34700840 bp on chr 3, affecting GW and GLWR in BC1F1s in NN, were mapped in the same region of qTGW3 that negatively modulates rice grain size and weight (Ying et al., 2018). Furthermore, qGW5, qGLWR5.1, and qTGW5 located in 4578959–5986015 bp on chr 5, affecting GW in BC1F1s under both locations, were mapped in the same region with the known gene GW5 that controls the width and weight of rice grain (Weng et al., 2008; Liu et al., 2017). qTGW4 located in 4578959–5986015 bp on chr 4, affecting TGW in BC1F1s in HF, was mapped with Gnp4, YDA1, and GIFI genes that act as positive regulators to enhance rice grain size and weight (He et al., 2017; Zhang et al., 2018). Similarly, qGL6.2 in the region of 26554537–27705641 bp on chr 6, affecting the GL in BC1F1s and HMPs in HF and RILs and BC1F1s in NN was mapped in the same region with GL6 and GW6a. GL6 positively regulates the grain length and weight through stimulating the cell propagation in grains (Wang et al., 2019a), and GW6a also controls the rice grain traits (Song et al., 2015; Shi et al., 2020; Zhang et al., 2020). GL6 and GW6a were mapped with qGL6.2, qGLWR6.3, and qGLWR6.4 with ID, OD, and UD heterotic effects. These two genes may have heterotic effects on grain traits. However, allelism between mQTL for grain shape, weight and yield described above and previously reported genes need to be confirmed further through gene cloning and fine-mapping.

Candidate genes for nine stable mQTL (Supplementary Table 8) were further inferred through bioinformatics and gene expression database. For example, qGYP1.2 contains four candidate genes, Os01g0218032, repressor of silencing 1a (ROS1a) gene essential for developing gametophytes (Ono et al., 2012). It was reported that any interference in ROS1a gene induced pollen and embryo sac defects in rice, consequently reduced the rice yield (Xu et al., 2020). Of the four candidate genes for qGYP2.3, only Os02g0671100 is mainly expressed in seeds and encodes OsFBDUF13 F-box and DUF domain containing protein (http://rice.uga.edu/cgi-bin/ORF_infoPage.cgi?orf=LOC_Os02g44990.1) that positively regulates the spikelet numbers (Kaur et al., 2018). Furthermore, F-box protein also reduced abiotic stress tolerance and promoted root growth in rice (Yan et al., 2011). For qGYP3.2, Os03g0715600 encodes EF-hand family protein (http://rice.uga.edu/cgi-bin/ORF_infoPage.cgi?orf=LOC_Os03g50760.1), reducing the yield loss by protecting the rice plant against biotic stress (Ye et al., 2017). Os03g0716400 (http://rice.uga.edu/cgi-bin/ORF_infoPage.cgi?orf=LOC_Os03g50790.1) and Os03g0719500 (http://rice.uga.edu/cgi-bin/ORF_infoPage.cgi?orf=LOC_Os03g51020.1) both expressed in seeds and encoded uncharacterized conserved proteins, so these both are likely candidate genes. Three candidate genes for qGLWR4.1, only Os04g0168100 encodes Zinc finger, C2H2-type domain containing protein (http://rice.uga.edu/cgi-bin/ORF_infoPage.cgi?orf=LOC_Os04g08600.1) that is master regulators of abiotic stress response in plants...
(Han et al., 2020), and particularly regulates salt and drought tolerance in rice (Huang et al., 2009). For qGL5.1 and qGLWR5.2, Os05g0569300 expressed in seeds and encodes G-box binding factor 1 (GBF1) (http://rice.uga.edu/cgi-bin/ORF_infopage.cgi?orf=LOC_Os05g49420.1), which play essential role in fast-growing pollen in higher plants (Richert et al., 2012). Two candidate genes for qGL6.1, qGW6.1, and qGLWR6.2, only Os06g0215200 (http://rice.uga.edu/cgi-bin/ORF_infopage.cgi?orf=LOC_Os06g11170.1) is mainly expressed in seeds, and encodes formin binding protein. Foramin rice morphology determinant (RMD) plays a significant role to determine morphology of rice. RMD mutants exhibit bending growth pattern in seedlings, and have aberrant panicle and seed shape. This study suggested that the rice formin protein plays a key role in assessing the morphology of plants through regulation of actin dynamics and accurate spatial organization of microtubules and microfilaments (Zhang et al., 2011; Chang et al., 2022).

Breeding technique aimed at grain quality and high yield in hybrid rice

The prime purpose of rice breeding is to attain good quality grain and yield (Xu et al., 2015). Owing to the low heritability and genotype-by-environment interaction, it becomes quite challenging for traditional breeding techniques to improve rice yield and quality traits simultaneously. Using knowledge of rice heterosis, it is sure to produce high yield and good quality hybrid rice varieties by introgressing or pyramiding FAs to achieve the heterosis, it is sure to produce high yield and good quality hybrid rice varieties by introgressing or pyramiding FAs to achieve the heterosis.

Specifying which QTL can be modified simultaneously for grain quality and yield traits is quite a challenge. The heterotic effect is likely to decrease as the heterotic advantage of a hybrid is reduced (Chang et al., 2022). This region also contains a GW5 region on Chr 5 (4.58 MB) containing qGW5, qTGW5, and qGY5 (Supplementary Table 6). This region also contains a known grain-size gene GW5 (Liu et al., 2017). The WSSM alleles at the detected QTL reduced GW and enhanced GY, suggesting that GW5 may improve grain size and yield of hybrid rice simultaneously. Likewise, it would be possible to increase grain yield and quality together by pyramiding FA with the same directions of gene effects for GY and grain size at the QTL regions 81315122–8339827 bp on chr 2, and 28238757–3003190 bp on chr 3, and 20001168–20792875 bp on chr 10, to attain best combinations of mQTL linked with quality and yield.

Additionally, in this research, 11 identified mQTL (qGL2, qGL4.1, qGL6.2, qGL11.1, qGW8, qGLWR2.2, qGLWR4.1, qGLWR9.3, qTGW9, qGY5.1, and qGY2.3) were found in RILs and BC1F1s, but not in HMPs, mainly contributed to BC1F1’s performance but with negligible heterotic effect (Supplementary Table 3). The same FA in WSSM and Q9311B at above 11 mQTL influenced BC1F1’s performance. Twenty-five mQTL found in both BC1F1s and HMPs contributed to BC1F1 performances and heterosis (Supplementary Table 3). Therefore, further, improvement can be achieved in hybrid rice yield and quality by eliminating UD loci and pyramiding OD/D loci that are polymorphic between the parents. Additionally, the FA found in this study will provide a valuable reference for future breeding activities to simultaneously improve grain quality and yield.

Two subspecies offer noticeable differences in grain shape. The geng-type alleles (qG5S5/G5W5, OsSPL16/GW8, GS6) increased the width of rice grain (Wang et al., 2012; Liu et al., 2017; Chang et al., 2022), while G5S and OsSPL13/GW7 of xian-type alleles increase grain length (Li et al., 2011; Si et al., 2016). In the current study, we found many candidate genes had different phenotypic effects on grain traits and GYBP between the two subspecies (Supplementary Table 8). So, in rice breeding for grain shape, weight and yield quality, we can transfer favorable geng alleles (haplotypes) of Os05g0567500, Os05g0569300, and Os01g0218032 into xian background to improve grain shape quality of xian variety. In a similar way, we can transfer favorable xian alleles of Os03g0715600, Os03g0716400, and Os03g0719500 into geng background to improve grain shape, weight, and grain yield quality of geng variety or pyramiding above non-allelic FA to simultaneously improve grain shape and weight quality as well as grain yield in rice breeding of xian-geng inter-subspecific heterosis.

Data availability statement

The original contributions presented in the study are publicly available. This data can be found here: https://rfgb.rmbreeding.cn/file/upload/rfgb/SNP%20data%20of%20RILs_frmXu_20220913_Read.xlsx.

Author contributions

HY, SZ, KC, CS, HW, and FjZ performed the experiments. SZ and HY drafted the manuscript. FZ and HY did data analysis. JX and CZ designed the experiments and revised the manuscript. All authors contributed to the article and approved the submitted version.

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Conflict of interest

HW and FjZ are employed by Hunan Tao-Hua-Yuan Agricultural Technologies Co. LTD, and CZ is employed by Win-All Hi-Tech Seed Co., Ltd., Hefei, China.
The remaining 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/fpls.2022.977349/full#supplementary-material
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