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Whole genome resequencing identifies candidate genes and allelic diagnostic markers for resistance to *Ralstonia solanacearum* infection in cultivated peanut (*Arachis hypogaea* L.)

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Bacterial wilt disease (BWD), caused by Ralstonia solanacearum is a major challenge for peanut production in China and significantly affects global peanut field productivity. It is imperative to identify genetic loci and putative genes controlling resistance to R. solanacearum (RRS). Therefore, a sequencingbased trait mapping approach termed "QTL-seq" was applied to a recombination inbred line population of 581 individuals from the cross of Yueyou 92 (resistant) and Xinhuixiaoli (susceptible). A total of 381,642 homozygous single nucleotide polymorphisms (SNPs) and 98,918 InDels were identified through whole genome resequencing of resistant and susceptible parents for RRS. Using QTL-seq analysis, a candidate genomic region comprising of 7.2 Mb (1.8–9.0 Mb) was identified on chromosome 12 which was found to be significantly associated with RRS based on combined Euclidean Distance (ED) and SNP-index methods. This candidate genomic region had 180 nonsynonymous SNPs and 14 InDels that affected 75 and 11 putative candidate genes, respectively. Finally, eight nucleotide binding site leucine rich repeat (NBS-LRR) putative resistant genes were identified as the important candidate genes with high confidence. Two diagnostic SNP markers were validated and revealed high phenotypic variation in the different resistant and susceptible RIL lines. These findings advocate the expediency of the QTLseq approach for precise and rapid identification of candidate genomic regions, and the development of diagnostic markers that are applicable in breeding disease-resistant peanut varieties.

KEYWORDS

peanut, resistance to *Ralstonia solanacearum*, QTL-seq analysis, candidate genes, diagnostic markers

1 Introduction

Bacterial wilt that is caused by Ralstonia solanacearum (R. solanacearum), is the most damaging bacterial disease that globally affects over 50 and 450 botanical families and plant species, respectively, including several economically important crops such as tobacco, peanut, tomato, and pepper (Salanoubat et al., 2002; Zhang et al., 2017). R. solanacearum is a free-living saprophyte that endures in soil and aquatic habitats for long durations (Genin and Boucher, 2004). R. solanacearum mostly infects plant roots, propagates in the xylem, disseminates into the stem, and then to the entire plant resulting in wilt and eventual death (Schell, 2000). Bacterial wilt disease often significantly reduces, by 10~30%, the yield and quality of peanut and other important crops; it may also result in complete yield loss (Zhang et al., 2017). Currently, no effective pesticide and biological control method exists to control this pathogen because of its wide host range and durable survival ability (Yu et al., 2011). Nevertheless, cultivating crop varieties that are genetically resistant has efficiently controlled this disease (Sunkara et al., 2014; Reddy, 2016), leading to the development and release of many resistant varieties of the peanut. However, there exists a looming threat of a breakdown of this genetic resistance in China, due to similar resistance mechanisms in both the cultivated varieties (such as Xiekangqing, Taishan Zhenzhu) and wild species (Janila et al., 2016; Luo et al., 2020). Despite the current search for varieties whose resistance is conferred via alternate mechanisms, it is imperative to determine the genomic regions and genes that encode resistance to augment the development of new varieties via genomics-assisted breeding (GAB) (Pandey et al., 2020).

Through analysis of various plant genomes, map-based cloning of plant genes that confer resistance to *R. solanacearum* was conducted for a few crop species. In *Arabidopsis*, a recessive *RRS1-R* encoding a Tir-NBS-LRR resistant protein with a WRKY domain in resistant line Nd-1 was first identified and cloned by fine mapping (Deslandes et al., 2003). It conferred resistance to GMI1000 when transferred into

Col-5 variants with the dominant susceptible allele. Furthermore, another resistant gene RPS4 was in the reverse orientation and directly upstream of RRS1-R. This physical association triggered host resistance to the pathogen (Narusaka et al., 2009). Relatedly, RRS1-R associated with RPS4 is a dimer that recognizes PopP2 of R. solanacearum to trigger RRS (Narusaka et al., 2014). In Arabidopsis, three quantitative trait loci (QTLs) for RRS were identified in 100 F9 recombinant inbred lines (RILs) from another cross of Col-0 \times Ler (Godiard et al., 2003). A putative leucine-rich repeat receptor-like kinase (LRR-RLK) gene named ERECTA was cloned and found to trigger RRS (Godiard et al., 2003). Recently, in peanut, two genes AhRRS5 and AhRLK1 (also known as AhCLAVATA1), encoding an NBS-LRR resistance protein and a receptor-like protein kinase, respectively, were identified by reverse genetics. Transgenic tobaccos that overexpressed these two genes conferred a significantly increased level of resistance to RRS, indicating that both R genes and RLKs are involved in resistance mechanisms against BWD (Zhang et al., 2017; Zhang et al., 2019). Hitherto, no resistance genes from other plants have been cloned and characterized by the map-based method.

Recently, several QTLs associated with RRS were effectively identified by QTL mapping in many crop species, including tomato (Thoquet et al., 1996; Carmeille et al., 2006; Wang et al., 2013; Shin et al., 2020), pepper (Mimura et al., 2009) (Du et al., 2019), potato (Habe et al., 2019), eggplant (Lebeau et al., 2013; Salgon et al., 2017), tobacco (Wang et al., 2013) and Medicago truncatula (Ben et al., 2013). Up to now, both sequencing-based trait mapping and gene discovery techniques are highly utilized due to low sequencing costs and the development of new methods that elucidate genomic loci and candidate genes associated with specific traits (Varshney et al., 2019; Pandey et al., 2020). Such efforts facilitate faster development of diagnostic markers which can be employed in GAB to accelerate the development of new peanut varieties (Pandey et al., 2020). In peanut, Zhao et al. (Zhao et al., 2016) first reported mapping QTL for RRS on the B02 chromosome using a moderately dense linkage map of 237 SSR and SNP markers. By combining restriction-site-associated DNA sequencing (RADseq) and bulk segregant analysis (BSA) techniques, they developed resistant-related SNP markers from the RIL population of crosses between resistant (Yueyou 92) and susceptible (Xinhuixiaoli) varieties. The two detected QTLs (qBW-1 and qBW-2) in the aforementioned RRS study accounted for 21% and 12% of the resistance phenotypic variance in the F₂ generation, respectively. Only two side-byside QTLs were found at the qBW-1 locus on the B02 chromosome in the F₈ generation. The resistant resource of Yueyou 92 was from a Chinese landrace Xiekangqing, which is a major source of parental types used for breeding BWD-resistant variants in South China (Janila et al., 2016; Luo et al., 2020).

The rapid QTL-seq approach is critical for identifying genomic regions of a trait of interest in plants and identifies QTLs based on BSA and next-generation sequencing (Takagi et al., 2013). QTL-seq was the preferred choice of a fast and effective method that identifies and maps QTLs of target traits in crop plants (Takagi et al., 2013). For example, it was to identify QTLs of the target trait in rice (Arikit et al., 2019; Bommisetty et al., 2020; Lei et al., 2020; Yang et al., 2021), cucumber (Lu et al., 2014; Cao et al., 2021; Zhang et al., 2021), chickpea (Das et al., 2014; Singh et al., 2016; Srivastava et al., 2017), tomato (Illa et al., 2015; Topcu et al., 2021), oilseed rape (Wang et al., 2016; Tudor et al., 2020; Dong et al., 2021), maize (Chen et al., 2018; Wang et al., 2021), and peanut (Pandey et al., 2017; Clevenger et al., 2018; Luo et al., 2019; Kumar et al., 2020; Luo et al., 2020; Topcu et al., 2021). In peanut, it was used to map genomic loci and candidate genes for the development of diagnostic markers for RRS in 195 RILs obtained by crossing Yuanza 9102 and Xuzhou 68-4 (Luo et al., 2019). A major and stable QTL (qBWRB02.1) on chromosome B02 was identified, which was significantly associated with RRS in three environments. Moreover, two SNP sites were confirmed in diverse breeding lines and cultivars. Unlike Yueyou 92, Yuanza 9102 was derived from the wild species Arachis diogoi that was resistant to BWD (Janila et al., 2016; Luo et al., 2019). A stable QTL for RRS was finely mapped via both linkage mapping and QTL-seq tools in a resistant peanut cultivar (Luo et al., 2020). Two hundred and sixty-eight RILs were sequenced, and the phenotypes of variants from the cross between Xuhua 13 (susceptible) and Zhonghua 6 (resistant) among five environments were evaluated. Using both SSR- and SNP-based genetic maps, the QTL qBWRB02-1 was identified on chromosome B02 as previously reported (Zhao et al., 2016), and this accounted for 37.79-78.86% phenotypic variation across the five environments. Two adjacent candidate QTL regions in the qBWRB02-1 locus were segmented into qBWRB02-1-1 (2.81-4.24 Mb) and qBWRB02-1-2 (6.54-8.75 Mb) (Luo et al., 2020). QBWRB02-1-1 accounted for 49.43-68.86% phenotypic variation explained (PVE), which was higher than that for gBWRB02-1-2 (3.96-6.48% PVE). Moreover, this was validated by competitive allele-specific PCR (KASP)

markers in different RILs and natural populations (Luo et al., 2020).

In this study, we utilized a QTL-seq approach to identify concomitant genomic regions, candidate resistance genes and diagnostic markers in a bacterial wilt-resistant peanut variety, Yueyou92. A 7.2 Mb candidate genomic region was elucidated on chromosome 12 significantly associated with RRS. This study reports successful discovery of followed by candidate resistance genes and validated markers for potential use in marker-assisted selection (MAS) for RRS in peanut breeding programs.

2 Materials and methods

2.1 Plant material and growth

Yueyou 92 (YY92), a variety that is highly resistant to BWD, was bred by the Guangdong Academy of Agricultural Sciences, China. It stemmed from Xiekangqing, which was resistant to *R. solanacearum* strains from different parts of China. In comparison, Xinhuixiaoli (XHXL) was a Chinese landrace that was highly susceptible to BWD. Their resistance validation was stable during multiple years of field assessment (Figure 1). A RIL population containing 581 lines was developed from the cross Yueyou 92 × Xinhuixiaoli using the single seed descent (SSD) method. A total of 581 F_{13} RILs were used for trait mapping for RRS. To assess the diagnostic markers, we utilized 18 resistant and 18 susceptible RILs for genotyping using allele-specific markers. All the RILs and parents were cultivated in a field in Yangzhong County (Sanming, Fujian, China).

2.2 Pathogen inoculation and resistance phenotyping

The 581 RILs were evaluated for RRS in three independent crop seasons i.e., in 2016 spring and autumn (2016S and 2016A) and in 2017 spring (2017S). The RILs of F_{11} , F_{12} , and F_{13} generations and parents were cultivated in two-row plots with 20 seeds in each season. One-month to 40-day-old RILs seedlings were inoculated via a previously described artificial method (Zhao et al., 2016; Zhang et al., 2017). Resistance phenotyping of the RILs was performed 25 days after inoculation in the different seasons. Disease symptoms were classified into six disease severity ratings (Figure 1C): (0) =the inoculated leaflets either remained green or were yellow at inoculating sites, but the entire plant was intact and lacked wilt; (1) = the inoculated leaflets had either wilted or fallen off, but the entire plant was intact and lacked wilt; (2) = the main stem/branches of the inoculated leaves had wilt and chlorosis; (3) = the leaves of non-inoculated branches had wilt or were faded green, but the main stem was green; (4) = the entire plant had wilted and died, and all its branches were greenish; and (5) =



the entire plant had wilted, dried, and was brownish. The disease index (DI) was calculated using the following formula:

Disease index =
$$\frac{\sum_{0}^{5} x_i y_i}{x_{max} \sum y_i} \times 100 \%$$

Where, x_i : disease grade value, x_{max} : the highest disease grade value, and y_i : the number of diseased plants corresponding to the disease rating.

The average DI was calculated for the three replications in a single environment. Statistical analysis of variance (ANOVA) was performed using the DPS7.5 software (Date Processing System, Science Press, China), Values are expressed as the mean \pm standard deviation or standard error as indicated.

Differences between groups were evaluated using one-way ANOVA. Statistical significance was set at P<0.05.

2.3 Extreme bulks construction and whole genome resequencing

The average DI for each RIL was calculated based on phenotyping data from the 2016S, 2016A, and 2017S seasons. We selected 30 resistant and 30 susceptible lines to construct the extreme R/S pool. To develop the resistant bulk (R-Bulk) for RRS, we selected 30 RILs with a low mean disease index and pooled the same amount of DNA from each into one. Similarly, DNA samples of 30 RILs with a high mean disease index were pooled to construct the susceptible bulk (S-Bulk) for RRS. The genomic DNA of these two extreme pools and those of the two parents was used to construct DNA sequencing libraries. Pairedend reads (151 bp) of four libraries were generated *via* the Illumina HiSeq 2500 platform (Illumina, Inc., USA) with a sequencing depth of approximately $30\times$ of the cultivated peanut genome (~2.7 Gb) for each pool and about $40\times$ for parental plants. The raw sequencing data of the four libraries have been deposited in the NCBI Sequence Read Archive (SRA) under the BioProject ID PRJNA851221.

2.4 SNP/InDel genotype detection and annotation

A QTL-seq approach was used to identify the QTLs for RRS (Figure S1) (Takagi et al., 2013). The quality of re-sequenced raw reads from the four libraries was checked. Low-quality reads (those with a proportion of uncalled bases >5%) and adapter sequences were culled. High-quality reads (those with more than 95% nucleotide base calls and high Phred quality scores) were aligned and mapped to the reference genome using the BWA software package (http://bio-bwa.sourceforge.net/) with the default parameters (Li and Durbin, 2009). For further analysis, the genome sequences of allotetraploid progenitors of the cultivated peanut Arachis hypogaea (Shitouqi) were downloaded from the Peanut Genome Resource website (http://peanutgr.fafu.edu.cn/) and used as the reference sequences. Duplicated reads were identified and filtered using Picard (http://broadinstitute.github.io/picard/) after mapping the clean reads to the reference genome. To determine the locations and effects of the SNP/InDel variants, we detected and filtered variants in the four libraries using the Genome Analysis Toolkit (GATK, https://software.broadinstitute.org/ gatk/) (Cingolani et al., 2012) and annotated using SnpEff software (V.5.0e; https://pcingola.github.io/SnpEff/) (Reumers et al., 2012).

2.5 Identification of candidate genomic regions

To identify the candidate genomic regions associated with RRS, we further filtered high-quality reads from S-bulk and R-bulk libraries by removing unpaired reads. To equalize the number of reads from each bulk, the filtered reads of the R-bulk were randomly reduced to the same number of the filtered reads of the S-bulk. During the analysis, the SNP sites with genotypes that differed between the two bulks were used to calculate both the sequencing depth of each base in the different bulks and the

Euclidean Distance (ED) value of each site. To eliminate the background noise, the original ED value was processed by power. In this study, the fifth power of the original ED was taken as the correlation value to eliminate the background noise. Then the distance method was used to fit the ED value. For every SNP and InDel in each bulk, ED values were calculated with the formula:

$$ED = \sqrt{(A_{R-bulk} - A_{S-bulk})^2 + (C_{R-bulk} - C_{S-bulk})^2 + (G_{R-bulk} - G_{S-bulk})^2 + (T_{R-bulk} - T_{S-bulk})^2}$$

Each A, G, C, and T letter represented the frequency of its corresponding DNA nucleotide in the resistant and susceptible bulks. The higher the ED value, the stronger the association between the variant with the target characteristic.

The SNP index value was calculated as follows:

$$SNP - index(aa) = \frac{Maa}{Maa + Paa}$$
$$SNP - index(ab) = \frac{Mab}{Mab + Pab}$$

 Δ SNP-index = SNP-index (aa)- SNP-index (ab)

 Δ SNP-index was calculated by subtracting the SNP-index of the R-bulk from the SNP-index of the S-bulk. SNP-index plots were generated using sliding window analysis with a window size of 2 Mb and increments of 50 Kb. The SNPs with SNP-index <0.3 or read depth<10 in both bulks were culled (Supplementary Figure 1). The SNP index of remaining SNPs as calculated from each bulk was physically plotted onto the 20 cultivated peanut chromosomes. ASNP index was calculated by subtracting the SNP index of the resistant bulk from the SNP index of the susceptible bulk. Notably, only those SNPs that had homozygous alleles in both bulks were selected for Δ SNP index calculation. Furthermore, SNP positions were considered as the causal SNPs responsible for the trait of interest if they passed the criterion Δ SNP index = -1. Δ SNP index = -1 indicated that the allele called in resistant bulk was the same as that of the resistant parent while an alternate base was called in susceptible bulk (Supplementary Figure 3). This analysis was also used in the InDel correlation analysis (Supplementary Figure 4). The DISTANCE method was used to fit Δ SNP-indexes and ED values, and the regions above the correlation threshold value (add value) were selected as those related to traits.

The candidate genomic regions related to RRS had the following significant Δ SNP/InDel index requirements: Δ SNP/ InDel index significantly deviated from the statistical confidence intervals under the null hypothesis of no QTLs at a *P* < 0.01 level, and SNP/InDel-index significantly deviated from 0.5 in both bulks. Moreover, the ED values for SNP and InDel were remarkably higher than 0.29 and 0.28, respectively. Finally, the two sets of genomic regions identified from the S-bulk and R-bulk assemblies were combined and considered the genomic regions associated with RRS.

2.6 Diagnostic marker development and validation

To validate the identified genomic regions for RRS, SNPs with different alleles in both bulks and near the intersection terminal were identified and a special marker was developed to narrow the candidate region. For the RIL lines under *R. solanacearum* treatments and with the highest and lowest DI values, we randomly selected 18 of each of these two groups, then extracted DNA from them as well as the parents and the other selected samples. The total volume for the PCR reaction was 20 µl, comprising DNA template: 50 ng, 2×PCR Master Mix: 5 µl, forward primer: 10 µM, and reverse primer: 10 µM. The PCR cycling conditions were as follows: 94°C, 3 min; 30–35 cycles of 94°C, 30 s; 56°C, 30 s; 72°C, 30 s; final extension at 72°C, 10 min. After PCR amplification, the targeted amplicons were identified *via* 1.2% agarose gel electrophoresis.

3 Results

3.1 Phenotype diversity and construction of extreme RRS bulks

To investigate variation in RRS levels of cultivated peanuts, we utilized the resistant "Yueyou 92" (RP) and susceptible "Xinhuixiaoli" (SP) varieties as parents to create multiple generations of segregating RILs populations (Figure 1A). The resistance rate was evaluated based on the severity of R. solanacearum infections in RILs, which was calculated as a disease index (DI). The DI value of Yueyou 92 was significantly lower than that of Xinhuixiaoli in three consecutive crop seasons (Figure 1A and Supplementary Table 1). The RILs population had a wide segregation of phenotype variations that formed two peaks of resistance distributions, displaying the main QTLs for RRS regulation (Figure 1B and Supplementary Table 1). Based on the mean values of the disease index in the three field environments, the 30 RILs with the lowest disease index (10.22-20.00%) and the 30 RILs with the highest index (81.68-92.79%) were selected for construction of the resistant (R-bulk) and susceptible bulks (S-bulk) respectively (Figure 1D and Supplementary Table 1). Furthermore, phenotypic identification of R- and S-bulk resistance in the greenhouse was like that in the field environment (Supplementary Figure 1).

3.2 Genome sequencing and SNP/InDel discovery and evaluation

Whole genome sequencing of the parents and the bulks DNA samples was performed on the Illumina HiSeq platform. A total of 114.67 and 103.10 Gb reads were generated for Yueyou 92 and Xinhuixiaoli, and 108.38 and 96.92 Gb for R-bulk and Sbulk respectively. Approximately 97.69% of the reads correctly mapped to the cultivated peanut cv. Shitouqi reference genome (Table 1). An average coverage depth of $42\times$ and $37\times$ of the reference genome was achieved by Yueyou 92 and Xinhuixiaoli reads respectively, and $38\times$ and $34\times$ depth for the R-bulk and Sbulk, respectively (Table 1). The mapping results showed that the genome was evenly covered, indicating that the sequencing randomness was good (Supplementary Figure 2).

SNPs/InDels were detected and extracted by the GATK software package. A total of 585,258 SNPs and 167,249 InDels were detected between two parents and 126,900 SNPs and 46,013 InDels were detected between the extreme pools, respectively. The occurrence of the SNPs was 3.5 times more than that of the InDels (Supplementary Table 2; Supplementary Figure 3). After filtering, 381,642 and 98,918 high-quality and homozygous SNP and InDel sites were respectively obtained (Supplementary Table S3). Based on the annotations, 72.7% and 55.4% of the SNPs and InDels, respectively, were in the intergenic region between the extreme pools. Approximately 15% of SNPs and 25% of InDels were upstream and downstream of genes, ~10% of variants in introns, and only ~3% of SNPs and 2.6% of InDels were in the coding region of the two bulks. About 34.6% and 54.7% of the SNPs in the CDS region caused synonymous and nonsynonymous coding variants, respectively. Similarly, Approximately 22.0% of the InDels in the CDS region caused frameshift variants (Supplementary Tables 2, 3).

3.3 Candidate genomic regions for RRS

Using the genome sequences of Arachis hypogaea (Shitouqi) as reference, the Euclidean distance (ED) and SNP index, including the Δ SNP-index, were calculated for each genomewide high-quality SNPs, from RP and SP (Figure 2; Supplementary Figure 6; Supplementary Figure 8). Then, candidate genomic regions for RRS were identified based on ED and Δ SNP-index plots through sliding window analysis of deviations from the threshold value at a 99% confidence level. By using an ED association algorithm, a major peak on Chr12 was identified for RRS, spanning 0-15.19 Mb with an ED > 0.29 (P<0.01) for the SNP. A 6.40 Mb (0.77-7.17 Mb) interval on Chr02 was also identified. By SNP-index and Δ SNP-index, only a genomic interval of 5.83 Mb (4.16-9.99 Mb) on Chr12 deviated from the threshold with the confidence level of *P*<0.01 (Figure 2; Supplementary Figure 7; Supplementary Figure 9), indicating the interval on Chr12 as the main region controlling the RRS. Moreover, the ED and InDel-indexes (referring to principles of Δ SNP-index) for each identified genomic InDel were calculated for RP and SP. The regions of similarity were confirmed at intervals of 0-7.0 Mb and 0-15 Mb on Chr02 and Chr12, respectively, for ED mapping and 7.49-9.99 Mb on Chr12 for

Sample ID	Genotype/ bulks	Total_reads	Clean reads	Clean_Base	Q30 (%)	GC (%)	Average depth(X)	Genome coverage ration_1X (%)	Genome coverage ration_5X (%)	Genome coverage ration_10X (%)	Mapped (%)	Properly_mapped (%)
RP	Resistant parent	467,742,004	382,713,001	114,670,878,822	89.89%	36.08%	42	97.72%	97.01%	96.42%	99.55%	96.59%
SP	Susceptible parent	602,985,958	344,100,662	103,101,745,626	90.68%	35.86%	37	97.42%	96.62%	95.77%	99.72%	96.31%
R-bulk	Resistant bulk	602,985,958	361,696,407	108,373,962,086	90.94%	35.99%	38	97.83%	97.11%	96.34%	99.74%	96.23%
S-bulk	Susceptible bulk	630,443,472	323,474,233	96,921,445,922	91.27%	35.94%	34	97.77%	96.99%	95.98%	99.74%	96.84%
The short r	eads of parents and	d the extreme bulks	were aligned t	o the genome seque	nces of cu	ltivated pe	anut, cv. Shitouc	ii, Arachis hypogaea Linn (Pe	anut Genome Resource: http:/	//peanutgr.fafu.edu.cn/).		

InDel-index association (Supplementary Figures 6-9). Once more, the candidate region on Chr12 was robust with a P<0.01 confidence level for both methods. As SNP-indexes enabled fine mapping, the 7.2 Mb (1.8–9.0 Mb) and 5.83 Mb (4.16–9.99 Mb) intervals on Chr12 were collectively identified as candidate region associated with RRS, at 95% and 99% confidence levels, respectively (Figure 3).

3.4 Genetic confirmation of candidate genomic region

To confirm the candidate genomic regions associated with RRS by the QTL-seq approach, we remapped the linkage group (LG) of the existing genetic map to the previously published and newly developed SNP markers. The QTL map had two QTLs located in LG1 (ChrB02) and LG10 in F2, which explained 21% and 12% of phenotype variations, and one QTL with two adjoining peaks in LG1 (ChrB02) in F8 (Zhao et al., 2016). The QTL on ChrB02 was located between SNP markers SNP79 and SNP129 in LG1, for which the LOD value was 3.91 and over 6.22, respectively (Zhao et al., 2016). We remapped the SNP markers to the cultivated peanut reference genome. The SNP79 and SNP129 markers were at 1.2 Mb and 9.2 Mb on Chr12, respectively, corroborating identified candidate genomic region through QTL-seq approach (Supplementary Figure 10). Recently, QTL analyses based on SLAF-seq were conducted to detect the candidate QTL region that confers RRS, and the concomitant genotyping and phenotyping data was used for mapping. This resulted in the identification of a consistent region between the SNP marker loci Marker7969064 and Marker7795914 (unpublished data), which explained 45% of the phenotype variations. Marker7969064 and Marke7795914 were located at 6.2 Mb and 8.7 Mb on Chr12, respectively (Supplementary Figure 10). The candidate genomic region associated with RRS as per QTL mapping corroborated that from the QTL-seq method. These results supported QTL-seq results, which revealed the candidate genomic region is associated with RRS.

3.5 Candidate genes associated with RRS

To narrow down the genomic regions and validate effective SNPs associated with RRS on Chr12, we selected a genomic region spanning 7.2 Mb on Chr12 and with 1807 effective SNPs and 629 InDels. Function annotation analysis of the 1807 SNPs revealed 503 intergenic SNPs; 461 intronic; 357 and 225 that were upstream and downstream of genes, respectively; two, six, and eight in 5' UTR, 3' UTR, and splice site regions, respectively; and 67 synonymous and 180 nonsynonymous (two resulted in stop codons) (Supplementary Table 4). Notably, 22 genes with nonsynonymous SNPs were predicted to encode for the NBS-

TABLE 1 Summary of whole genome re-sequencing of parents and bulk lines for bacterial wilt resistance.



LRR type disease resistance proteins, including AH12G01510, AH12G01540, AH12G01550, AH12G01560, AH12G01570, AH12G01600, AH12G01900, AH12G01920, AH12G01980, AH12G02020, AH12G02090, AH12G02120, AH12G02130, AH12G02310, AH12G02330, AH12G02370, AH12G02390, AH12G02410. AH12G02880, AH12G03230, AH12G03600 and AH12G06320 (Table 2). The AH12G01460 and AH12G06300 encode a receptor-like kinase protein and exocyst subunit Exo70 family protein B2 subunit respectively. AH12G03290 and AH12G05320 both encode Serine/threonine-protein phosphatase 7. The other putative candidate genes encoded various kinds of proteins (Table 3). Notably, eight of 22 candidate NBS-LRR resistant genes were identified with high confidence as important candidate genes with the Δ SNP values above 0.60 (Figure 4). Moreover, among the 629 InDels, 152 and 189 were in the intergenic and intronic regions, respectively, 146 and 114 were upstream and downstream of genes, respectively, five and four in 5' UTR and 3' UTR regions, respectively, nine and five resulted in frame shifts and codon changes respectively (Supplementary Table 5). The 180 nonsynonymous SNPs affected 75 putative candidate genes

associated with RRS (Table 2), whereas 14 InDels affected 11 genes (Table 3). Among them, six in NBS-LRR genes *AH12G01920*, *AH12G01980*, *AH12G02090*, *AH12G02390*, *AH12G02440*, *AH12G02600* affected the encoded functions as well as Δ SNP-index results (Table 3). Taken together, these results support the hypothesis that the six NBS-LRR resistance genes might act as the candidate genes related to RRS.

3.6 Allele-specific marker development and validation

To evaluate the specificity of the allele marker of the resistant and susceptible peanut cultivars, we targeted 44 SNPs from candidate NBS-LRR genes for RRS for the development of allelespecific markers (Supplementary Table 6). Allele-specific primers for 44 SNPs were successfully generated. All 44 allelespecific markers were checked for polymorphisms between parental genotypes of the RIL population (YY92 and XHXL). Of the 44 markers, 30 allele markers had good amplification, whereas 14 markers did not yield amplicons with clear bands



sliding-window analyses of cultivated peanut chromosomes. (E) Δ SNP-index plot generated by using the Shitouqi assembly as a reference genome. Label definitions from outside to inside: upper probability values at 99% (orange) and 95% (green) confidence levels. Δ SNP-index, lower probability values at 95% (green), and 99% (orange) confidence levels. (F, G) InDel-index of R- and S-bulks plots generated by sliding-window analyses of cultivated peanut chromosomes. (H) Δ InDel-index plot generated by using the Shitouqi assembly as a reference genome. Label definitions from outside to inside: upper probability values at 99% (orange) and 95% (green) confidence levels. Δ SNP-index, probability values at 95% (green) and 99% (orange) confidence levels.

from samples of parental genotypes. Of the 30 amplified markers, two markers (qRRS18 and qRRS19) in AH12G03230 and AH12G06320 genes co-segregated with RRS and may thus be deployed for RRS breeding (Figure 5). These two polymorphic markers were validated on a panel of diverse genotypes containing the resistant parent (Yueyou 92), 18 introgression-resistant RIL lines (YX131, YX189, YX284, YX303, YX636, YX712, YX759, YX905, YX962, YX540, YX544, YX793, YX802, YX875, R160, R201, R215 and R739), the susceptible parent (Xinhuixiaoli), and 18 susceptible RIL lines (YX32, YX68, YX160, YX211, YX293, YX554, YX622, YX840, YX178, R123, R592, YX57, YX80, YX95, YX299, YX469, YX707 and YX939). The primers for the diagnostic marker 'qRRS18' amplified a 302-bp fragment in the susceptible parent and different susceptible RIL lines, but none in the resistant genotypes (Figure 5A). In contrast, primers for another diagnostic marker 'qRRS19' amplified 217-bp fragment in the resistant genotypes and none in susceptible lines (Figure 5G). Most importantly, these two diagnostic markers (qRRS18 + qRRS19) could be further developed and used to distinguish between homozygotes and heterozygotes in the segregating population; i.e., susceptible lines will have a 302-bp allele from the marker 'qRRS18' and resistant lines will have a 217-bp allele from the marker 'qRRS19'. These two markers can be used as diagnostic marks for breeding resistant bacterial wilt varieties *via* MAS approach.

4 Discussion

With the advent of complete genome sequencing of diploid progenitor species and cultivated tetraploid variants, the QTL-

Gene ID	Chromosome	Physical position (bp)	Reference Genome	Alternative site	Resistant bulk (RB) base	Number of reads covering the site (X cov- erage) in resis- tant bulk (RB)	RB Depths of Ref, Alt	SNP- index of RB	Susceptible bulk (SB) base	Number of reads covering the site (X cov- erage) in sus- ceptible bulk (SB)	SB Depths of Ref, Alt	SNP- index of SB	Delta SNP- index (RB SNP- index-SB SNP- index)	SNP substitution effect	Amino acid change	U95 (95% confidence interval upper side)	L95 (95% confidence interval lower side)	U99 (99% confidence interval upper side)	L99 (99% confidence interval lower side)	Gene function
AH12G01450	Chr12	1817271	А	G	R	25	21,4	0.84	R	10	1,9	0.10	0.74	NON_SYNONYMOUS_CODING	Tct/Cct	0.459825	-0.460953	0.585964	-0.588628	AT-rich interactive domain-containing protein 1
AH12G01450	Chr12	1823428	С	Т	Y	29	25,4	0.86	Y	15	3,12	0.20	0.66	NON_SYNONYMOUS_CODING	Gaa/ Aaa	0.459812	-0.460941	0.585948	-0.588612	AT-rich interactive domain-containing protein 1
AH12G01460	Chr12	1832855	С	Т	Y	19	15,4	0.79	Y	10	1,9	0.10	0.69	NON_SYNONYMOUS_CODING	Gaa/ Aaa	0.4598	-0.460929	0.585932	-0.588596	Receptor-like protein B6:U6kinas 4
AH12G01510	Chr12	1864236	С	G	С	19	19,0	1.00	S	8	6,2	0.75	0.25	NON_SYNONYMOUS_CODING	Caa/ Gaa	0.459761	-0.460891	0.585885	-0.588548	Putative disease resistance RPP13- like protein 1
AH12G01540	Chr12	1888227	G	С	G	17	16,1	0.94	S	5	3,2	0.60	0.34	NON_SYNONYMOUS_CODING	Gga/ Cga	0.459735	-0.460866	0.585852	-0.588516	Putative disease resistance RPP13- like protein 1
AH12G01550	Chr12	1889601	G	А	G	19	19,0	1.00	R	5	2,3	0.40	0.60	NON_SYNONYMOUS_CODING	Gaa/ Aaa	0.459735	-0.460866	0.585852	-0.588516	Putative disease resistance RPP13- like protein 1
AH12G01550	Chr12	1889607	А	С	А	20	20,0	1.00	М	6	2,4	0.33	0.67	NON_SYNONYMOUS_CODING	Aat/Cat	0.459735	-0.460866	0.585852	-0.588516	Putative disease resistance RPP13- like protein 1
AH12G01550	Chr12	1889737	Т	А	Т	22	22,0	1.00	W	5	3,2	0.60	0.40	NON_SYNONYMOUS_CODING	gTt/gAt	0.459735	-0.460866	0.585852	-0.588516	Putative disease resistance RPP13- like protein 1
AH12G01560	Chr12	1889759	Т	А	Т	17	17,0	1.00	W	5	4,1	0.80	0.20	NON_SYNONYMOUS_CODING	gaT/ gaA	0.459735	-0.460866	0.585852	-0.588516	Putative disease resistance RPP13- like protein 1
AH12G01570	Chr12	1896280	А	G	R	21	19,2	0.90	R	5	2,3	0.40	0.50	NON_SYNONYMOUS_CODING	gAt/ gGt	0.459722	-0.460854	0.585837	-0.5885	Putative disease resistance RPP13- like protein 1
AH12G01570	Chr12	1896290	G	А	R	22	20,2	0.91	R	5	2,3	0.40	0.51	NON_SYNONYMOUS_CODING	atG/atA	0.459722	-0.460854	0.585837	-0.5885	Putative disease resistance RPP13- like protein 1
AH12G01570	Chr12	1896327	G	Т	G	17	16,1	0.94	К	4	1,3	0.25	0.69	NON_SYNONYMOUS_CODING	Ggc/ Tgc	0.459722	-0.460854	0.585837	-0.5885	Putative disease resistance RPP13- like protein 1
AH12G01570	Chr12	1896685	G	А	G	14	13,1	0.93	R	5	3,2	0.60	0.33	NON_SYNONYMOUS_CODING	cGa/ cAa	0.459722	-0.460854	0.585837	-0.5885	Putative disease resistance RPP13- like protein 1
AH12G01570	Chr12	1896707	С	А	С	16	15,1	0.94	М	5	3,2	0.60	0.34	NON_SYNONYMOUS_CODING	ttC/ttA	0.459722	-0.460854	0.585837	-0.5885	Putative disease resistance RPP13- like protein 1
AH12G01600	Chr12	1929770	С	G	С	23	23,0	1.00	S	11	3,8	0.27	0.73	NON_SYNONYMOUS_CODING	Ctt/Gtt	0.459683	-0.460815	0.585788	-0.588452	Putative disease resistance RPP13- like protein 1
AH12G01600	Chr12	1929827	Т	А	Т	22	22,0	1.00	W	6	2,4	0.33	0.67	NON_SYNONYMOUS_CODING	Tgt/Agt	0.459683	-0.460815	0.585788	-0.588452	Putative disease resistance RPP13- like protein 1

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	Gene function	Putative disease resistance RPP13- 'like protein 1	Putative disease resistance RPP13- like protein 1	Putative disease resistance RPP13- 'like protein 1	Putative disease resistance RPP13- like protein 1	Pentatricopeptide repeat-containing protein	HXXXD-type acyl- transferase family protein											
	1.99 (99% confidence interval lower side)	-0.588452	-0.588435	-0.588435	-0.588435	-0.588387	-0.588095	-0.588095	-0.588095	-0.588095	-0.588095	-0.588095	-0.588095	-0.588095	-0.588095	-0.587987	-0.587987	-0.587987
	U99 (99% confidence interval upper side)	0.585788	0.585771	0.585771	0.585771	0.585722	0.58544	0.58544	0.58544	0.58544	0.58544	0.58544	0.58544	0.58544	0.58544	0.585336	0.585336	0.585336
	1.95 (95% confidence interval lower side)	-0.460815	-0.460802	-0.460802	-0.460802	-0.460764	-0.460528	-0.460528	-0.460528	-0.460528	-0.460528	-0.460528	-0.460528	-0.460528	-0.460528	-0.46044	-0.46044	-0.46044
	U95 (95% confidence interval upper side)	0.459683	0.45967	0.45967	0.45967	0.45963	0.459384	0.459384	0.459384	0.459384	0.459384	0.459384	0.459384	0.459384	0.459384	0.459292	0.459292	0.459292
	Amino acid change	Tct/Gct	aAg/ aTg	ttA/ttC	tCg/tTg	aAa/ aGa	aaA/ aaC	Aca/ Cca	Ttt/Ctt	Aat/Gat	aAg/ aGg	Aat/Gat	ttC/ttG	Gtt/Att	atT/atG	agG/ agT	aGa/ aTa	gTa/ gCa
	SNP substitution effect	Non_synonymous_coding	NON_SYNONYMOUS_CODING	DNN_SVNONYMOUS_CODING	DNN_SVNONYMOUS_CODING	DNN_SVNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	DNIGOSUONYMOUY_CODING	DNN_SVNONYMOUS_CODING	DNN_SVNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	DNIGOSUONYMOUY_CODING	NON_SYNONYMOUS_CODING	DNN_SVNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING
	Delta SNP- index (RB SNP- index-SB SNP- SNP- index)	0.67	0.60	0.73	0.67	0.45	0.29	0.28	0.21	0.04	0.07	0.24	0.27	0.26	0.45	0.18	0.22	0.27
	SNP- index of SB	0.33	0.40	0.27	0.33	0.29	0.67	0.68	0.70	0.90	0.93	0.69	0.65	0.70	0.55	0.82	0.78	0.68
	SB Depths of Ref, Alt	2,4	2,3	3,8	3,6	4,10	14,7	13,6	16,7	18,2	13,1	11,5	13,7	14,6	6,5	9,2	7,2	15,7
	Number of reads covering the site (X cov- erage) in sus- ceptible bulk (SB)	Q	ю	11	9	14	21	19	23	20	14	16	20	20	11	Ξ	6	22
	Susceptible bulk (SB) base	К	M	М	Υ	Υ	М	М	Υ	R	В	К	s	R	K	К	K	Υ
	SNP- index of RB	1.00	1.00	1.00	1.00	0.74	960	0.96	160	0.94	1.00	0.93	0.92	0.96	1.00	1.00	1.00	0.95
	RB Depths of Ref, Alt	20,0	15,0	12,0	13,0	14,5	23,1	24,1	20,2	15,1	21,0	25,2	22,2	24,1	21,0	21,0	21,0	21,1
	Number of reads covering the site (X cov- erage) in resis- tant bulk (RB)	20	15	12	13	19	24	25	22	16	21	27	24	25	21	21	21	22
	e Resistant bulk (RB) base	H	A	A	C	Υ	A	A	Y	A	A	К	S	IJ	Ŧ	IJ	IJ	Н
	Alternative site	U	Ŧ	C	Т	C	C	U	U	U	U	U	G	A	U	Т	Т	O
	Reference Genome	Т	V	¥	C	F	A	V	г	A	V	A	U	G	F	G	G	н
	hysical osition (bp)	929839	931628	931686	931802	962973	019013	019053	019086	019242	019615	019677	019736	019773	019997	020027	020059	020110
Continued	Chromosome P	Chr12 15	Chr12 15	Chr12 15	Chr12 It	Chr12 15	Chr12 2(Chr12 2(Chr12 21	Chr12 21	Chr12 2(Chr12 2(Chr12 2(Chr12 21	Chr12 2(Chr12 21	Chr12 2(Chr12 20
TABLE 2	Gene ID	AH12G01600	AH12G01600	AH12G01600	AH12G01600	AH12G01630	AH12G01670											

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	Gene function	HXXXD-type acyl- transferase family protein	Subtilase family protein	Glycerol-3- phosphate 2-0- acyltransferase 6	UDP- glycosyltransferase 91A1	Putative disease resistance RPP13- like protein 1	UDP- glycosyltransferase 91A1	Putative disease resistance protein At3g14460	Putative disease resistance protein At3g14460	Putative disease resistance RPP13- like protein 1								
	1.99 (99% confidence interval lower side)	-0.587987	-0.587987	-0.587987	-0.587987	-0.587987	-0.587699	-0.58739	-0.586617	-0.585683	-0.585627	-0.585572	-0.585432	-0.585432	-0.585158	-0.585158	-0.585158	-0.585158
	U99 (99% confidence interval upper side)	0.585336	0.585336	0.585336	0.585336	0.585336	0.585059	0.584757	0.584017	0.583129	0.58308	0.58303	0.582905	0.582905	0.582663	0.582663	0.582663	0.582663
	1.95 (95% confidence interval lower side)	-0.46044	-0.46044	-0.46044	-0.46044	-0.46044	-0.460203	-0.459954	-0.459339	-0.45859	-0.458548	-0.458507	-0.458401	-0.458401	-0.458197	-0.458197	-0.458197	-0.458197
	U95 (95% confidence interval upper side)	0.459292	0.459292	0.459292	0.459292	0.459292	0.459047	0.458788	0.458144	0.457359	0.457315	0.457272	0.457162	0.457162	0.456951	0.456951	0.456951	0.456951
	Amino acid change	Gta/Ata	gaG/ gaC	Act/Gct	Atg/ Gtg	aGg/ aAg	Ttt/Ctt	Agt/ Ggt	aCa/ aTa	tCc/tTc	cAc/ cGc	Aga/ Gga	gGa/ gAa	tgG/ tgC	cCa/ cTa	cCt/cTt	Aga/ Gga	Ccc/ Tcc
	SNP substitution effect	NON_SYNONYMOUS_CODING	DNIGOD_SUONYAOUS_CODING	500 SYNONYAOUS_CODING	NON_SYNONYMOUS_CODING	DNIGOD_SUOMYNOUY_CODING	NON_SYNONYAOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYAOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYAOUS_CODING	NON_SYNONYAOUS_CODING	500 SYNONYAOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	5NIQOJ_SUOMYNONY_OUS_CODING	NON_SYNONYMOUS_CODING
	Delta SNP- index (RB SNP- index-SB SNP- SNP- index)	0.34	0.36	0.28	0.20	0.19	0.11	0.61	0.71	0.65	0.74	0.81	0.75	0.72	0.00	0.02	0.06	0.17
	SNP- index of SB	0.55	0.54	0.63	0.77	0.77	0.85	0.15	0.18	0.23	0.13	0.07	0.13	0.16	0.89	0.89	0.88	0.71
	SB Depths of Ref, Alt	17,14	15,13	19,11	20,6	20,6	17,3	2,11	3,14	3,10	2,13	1,13	2,13	3,16	17,2	16,2	15,2	10,4
	Number of reads covering the site (X cov- erage) in sus- ceptible bulk (SB)	31	28	30	26	26	20	13	17	13	15	14	15	19	19	18	17	14
	Susceptible bulk (SB) base	В	S	R	В	R	Υ	Υ	Y	Υ	R	К	R	S	Υ	Υ	R	Y
	SNP- index of RB	0.89	06.0	0.91	0.97	0.96	0.96	0.76	0.89	0.88	0.87	0.88	0.88	0.88	0.89	0.91	0.94	0.89
	RB Depths of Ref, Alt	24,3	27,3	30,3	30,1	25,1	24,1	13,4	16,2	23,3	20,3	22,3	23,3	21,3	17,2	20,2	15,1	16,2
	Number of reads covering the site (X cov- erage) in resis- tant bulk (RB)	27	30	33	31	26	25	17	18	26	23	25	26	24	19	22	16	18
	Resistant bulk (RB) base	К	S	A	¥	G	Т	Υ	U	Υ	В	ы	R	s	Υ	Y	Я	Y
	Alternative site	¥	C	IJ	IJ	V	U	C	F	Т	IJ	U	¥	U	H	H	IJ	Ŧ
	Reference Genome	(7)	(7)		~	(5)	L	2	0	0	7	~	(7)	(7)	0	0	7	0
	hysical osition (bp)	120205	120297	. 120334	120367	120380)54547	106320	249782	127316	132994	148354	473114	473124	534393	534405	534443	534503
Jued	pre pr	20	20	20	20	20	20	21	22	24	24	24	24	24	25	25	25	55
Contil	Chromo	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12
TABLE 2	Gene ID	AH12G01670	AH12G01670	AH12G01670	AH12G01670	AH12G01670	AH12G01690	AH12G01720	AH12G01780	AH12G01880	AH12G01900	AH12G01910	AH12G01920	AH12G01920	AH12G01980	AH12G01980	AH12G01980	AH12G01980

Image: sector in the	2	Continue	q																		
301 -		Chromosome	Physical position (bp)	Reference Genome	Alternative site	Resistant bulk (RB) base	Number of reads covering the site (X cov- erage) in resis- tant bulk (RB)	RB Depths of Ref, Alt	SNP- index of RB	Susceptible bulk (SB) base	Number of reads covering the site (X cov- erage) in sus- ceptible bulk (SB)	SB Depths of Ref, Alt	SNP- index of SB	Delta SNP- index (RB SNP- index-SB SNP- index)	SNP substitution effect	Amino acid change	U95 (95% confidence interval upper side)	L95 (95% confidence interval lower side)	U99 (99% confidence interval upper side)	L99 (99% confidence interval lower side)	Gene function
1 3111 1 31<	_	Chr12	2620504	A	C	A	25	25,0	1.00	М	6	7,2	0.78	0.22	Non_synonymous_coding	Tac/ Gac	0.456824	-0.458079	0.582535	-0.584984	Putative disease resistance RPP13- like protein 1
0 400 0	~	Chr12	2621513	C	Y	М	22	19,3	0.86	М	13	8,5	0.62	0.25	NON_SYNONYMOUS_CODING	aGg/ aTg	0.456824	-0.458079	0.582535	-0.584984	Putative disease resistance RPP13- like protein 1
0.1 0.11 1 1 0.1	~	Chr12	2641657	U	C	s	13	12,1	0.92	S	19	3,16	0.16	0.77	NON_SYNONYMOUS_CODING	gCc/ gGc	0.456806	-0.458062	0.582519	-0.584957	Putative disease resistance RPP13- like protein 1
0 310 1 0 1 0	~	Chr12	2651310	C	Т	Υ	13	6,7	0.54	Y	11	8,3	0.27	0.27	NON_SYNONYMOUS_CODING	gGt/ gAt	0.456798	-0.458054	0.582513	-0.584944	Putative disease resistance RPP13- like protein 1
010 010 0 <td></td> <td>Chr12</td> <td>2651350</td> <td>×</td> <td>C</td> <td>W</td> <td>1</td> <td>4,3</td> <td>0.43</td> <td>М</td> <td>6</td> <td>7,2</td> <td>0.22</td> <td>0.21</td> <td>NON_SYNONYMOUS_CODING</td> <td>Ttg/Gtg</td> <td>0.456798</td> <td>-0.458054</td> <td>0.582513</td> <td>-0.584944</td> <td>Putative disease resistance RPP13- like protein 1</td>		Chr12	2651350	×	C	W	1	4,3	0.43	М	6	7,2	0.22	0.21	NON_SYNONYMOUS_CODING	Ttg/Gtg	0.456798	-0.458054	0.582513	-0.584944	Putative disease resistance RPP13- like protein 1
0 0 1 1 1 0	0	Chr12	2651430	Т	¥	M	6	4,5	0.56	Т	10	9,1	0.10	0.46	NON_SYNONYMOUS_CODING	aAt/aTt	0.456798	-0.458054	0.582513	-0.584944	Putative disease resistance RPP13- like protein 1
0 0 1	0	Chr12	2777261	IJ	F	К	11	9,2	0.82	г	14	0,14	0.00	0.82	Non_synonymous_coding	agG/ agT	0.456652	-0.457914	0.582398	-0.584766	Putative disease resistance RPP13- like protein 1
0 0 1 0 1	0	Chr12	2816492	Т	IJ	К	11	13,4	0.76	К	15	2,13	0.13	0.63	TSOL_LOF_	tAa/tCa	0.456653	-0.457913	0.582412	-0.582412	Putative disease resistance RPP13- like protein 1
0 0 2 0 1 0 10	0	Chr12	2820701	IJ	¥	IJ	23	23,0	1.00	R	12	11,1	0.92	0.08	NON_SYNONYMOUS_CODING	aGg/ aAg	0.456665	-0.457925	0.582427	-0.584779	Putative disease resistance RPP13- like protein 1
0 0 0 0 1 0 1 0 1 0	0	Chr12	2820718	V	C	A	24	24,0	1.00	М	13	11,2	0.85	0.15	Non_synonymous_coding	Atc/Ctc	0.456665	-0.457925	0.582427	-0.584779	Putative disease resistance RPP13- like protein 1
0 Christian 223081 C A C A C A C A C A C A C A C A C <	0	Chr12	2820776	A	IJ	A	28	28,0	1.00	R	15	12,3	0.80	0.20	Non_synonymous_coding	gAa/ gGa	0.456665	-0.457925	0.582427	-0.584779	Putative disease resistance RPP13- like protein 1
0 Chrls 23033 C <thc< th=""> C C<</thc<>	0	Chr12	2820802	¥	C	¥	27	27,0	1.00	М	16	12,4	0.75	0.25	Non_synonymous_coding	Att/Ctt	0.456665	-0.457925	0.582427	-0.584779	Putative disease resistance RPP13- like protein 1
0 Ch12 232084 T A T 36 10 V 11 0.79 0.11 0.45795 0.45795 0.45795 0.58477 0.48705 0 Ch12 22094 G 1 1 56 0.45 0.45 0.45795 0.58475 Pastime ERP13- 0 Ch12 22094 G 20 10 R 11 56 0.45 0.45795 0.45795 0.58475 Pastime ERP13- 0 Ch12 22094 G 20 10 R 11 56 0.45 0.45795 0.58477 1 <td>0</td> <td>Chr12</td> <td>2820833</td> <td>IJ</td> <td>C</td> <td>IJ</td> <td>28</td> <td>28,0</td> <td>1.00</td> <td>S</td> <td>15</td> <td>11,4</td> <td>0.73</td> <td>0.27</td> <td>Non_synonymous_coding</td> <td>gGt/ gCt</td> <td>0.456665</td> <td>-0.457925</td> <td>0.582427</td> <td>-0.584779</td> <td>Putative disease resistance RPP13- like protein 1</td>	0	Chr12	2820833	IJ	C	IJ	28	28,0	1.00	S	15	11,4	0.73	0.27	Non_synonymous_coding	gGt/ gCt	0.456665	-0.457925	0.582427	-0.584779	Putative disease resistance RPP13- like protein 1
0 0 0 2 2 2 10 R 11 5,6 0.45 0.45 0.45725 0.58477 0.58477 0.44795 <td></td> <td>Chr12</td> <td>2820846</td> <td>Т</td> <td>¥</td> <td>Ţ</td> <td>26</td> <td>26,0</td> <td>1.00</td> <td>M</td> <td>14</td> <td>11,3</td> <td>0.79</td> <td>0.21</td> <td>NON_SYNONYMOUS_CODING</td> <td>agT/ agA</td> <td>0.456665</td> <td>-0.457925</td> <td>0.582427</td> <td>-0.584779</td> <td>Putative disease resistance RPP13- like protein 1</td>		Chr12	2820846	Т	¥	Ţ	26	26,0	1.00	M	14	11,3	0.79	0.21	NON_SYNONYMOUS_CODING	agT/ agA	0.456665	-0.457925	0.582427	-0.584779	Putative disease resistance RPP13- like protein 1
0 Chrl2 282456 A C M 23 0.87 M 23 4.19 0.17 0.70 NON_SYNONYMOUS_CODING 64/6665 0.457925 0.584.779 70843795 resistance RPP13- resistance RPP13- 1	~	Chr12	2820944	IJ	¥	IJ	22	22,0	1.00	R	11	5,6	0.45	0.55	Non_synonymous_coding	cGg/ cAg	0.456665	-0.457925	0.582427	-0.584779	Putative disease resistance RPP13- like protein 1
0 Chr12 2822465 A G R 21 18,3 0.86 R 22 5,17 0.23 0.63 NON_SYNONYMOUS_CODING cAUGG 0.456665 -0.457925 0.582427 -0.584779 Putative disease restature RPP13-	~	Chr12	2822450	¥	C	М	23	20,3	0.87	М	23	4,19	0.17	0.70	Non_synonymous_coding	gAt/gCt	0.456665	-0.457925	0.582427	-0.584779	Putative disease resistance RPP13- like protein 1
	0	Chr12	2822465	×	U	К	21	18,3	0.86	Ж	22	5,17	0.23	0.63	NON_SYNONYMOUS_CODING	cAt/cGt	0.456665	-0.457925	0.582427	-0.584779	Putative disease resistance RPP13- like protein 1

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	Gene function	Putative disease resistance RPP13- like protein 1																
	L99 (99% confidence interval lower side)	-0.584779	-0.584779	-0.584779	-0.584779	-0.5848	-0.5848	-0.5848	-0.5848	-0.5848	-0.5848	-0.5848	-0.5848	-0.5848	-0.5848	-0.584825	-0.584825	-0.584825
	U99 (99% confidence interval upper side)	0.582427	0.582427	0.582427	0.582427	0.582449	0.582449	0.582449	0.582449	0.582449	0.582449	0.582449	0.582449	0.582449	0.582449	0.582471	0.582471	0.582471
	1.95 (95% confidence interval lower side)	-0.457925	-0.457925	-0.457925	-0.457925	-0.457942	-0.457942	-0.457942	-0.457942	-0.457942	-0.457942	-0.457942	-0.457942	-0.457942	-0.457942	-0.457961	-0.457961	-0.457961
	U95 (95% confidence interval upper side)	0.456665	0.456665	0.456665	0.456665	0.456684	0.456684	0.456684	0.456684	0.456684	0.456684	0.456684	0.456684	0.456684	0.456684	0.456704	0.456704	0.456704
	Amino acid change	tAt/tGt	Gca/ Cca	Gaa/ Caa	Ctg/Atg	Ctt/Att	cGc/ cAc	aAg/ aGg	Tat/Cat	agG/ agT	Aga/ Gga	Tcc/ Acc	Caa/ Gaa	Aaa/ Gaa	cGc/ cTc	Aag/ Gag	Gac/ Cac	cCc/ cGc
	SNP substitution effect	Non_synonymous_coding	NON_SYNONYMOUS_CODING	DNN_SVNONYMOUS_CODING	DNN_SYNONYMOUS_CODING	500 SUDANONYAOUS_CODING	5NNON_SVNONYAOUS_CODING	NON_SYNONYMOUS_CODING	500 SUDANONYAOUS_CODING	DNN_SVNONYMOUS_CODING	9NINON_SUNNNYAOUS_CODING	DNN_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	9NIGO_SUONYMOUY_CODING	DNN_SVNONYMOUS_CODING	NON_SYNONYMOUS_CODING	DNN_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING
	Delta SNP- index (RB SNP- index-SB SNP- SNP- index)	0.67	0.71	0.80	0.83	0.53	0.01	-0.04	-0.03	0.44	0.48	0.42	0.44	0.20	0.60	0.75	0.89	0.72
	SNP- index of SB	0.33	0.29	0.20	0.17	0.21	0.92	0.96	0.96	0.56	0.52	0.58	0.56	0.80	0.33	0.25	0.05	0.12
	SB Depths of Ref, Alt	2,4	2,5	1,4	1,5	4,15	23,2	24,1	24,1	10,8	13,12	15,11	14,11	12,3	2,4	1,3	1,18	2,15
	Number of reads covering the site (X cov- erage) in sus- ceptible bulk (SB)	Q	2	ŝ	9	19	25	25	25	18	25	26	25	15	9	4	19	17
	Susceptible bulk (SB) base	Y	R	s	М	M	U	A	Т	К	ы	M	S	Ж	К	К	C	s
	SNP- index of RB	1.00	1.00	1.00	1.00	0.74	0.93	0.92	0.93	1.00	1.00	1.00	1.00	1.00	0.93	1.00	0.94	0.84
	RB Depths of Ref, Alt	19,0	20,0	26,0	24,0	14,5	25,2	24,2	25,2	22,0	17,0	15,0	16,0	11,0	14,1	21,0	17,1	21,4
	Number of reads covering the site (X cov- erage) in resis- tant bulk (RB)	19	20	26	24	19	27	26	27	22	11	15	16	Ξ	15	21	18	25
	Resistant bulk (RB) base	Ŧ	IJ	IJ	U	M	Я	R	Y	U	×	Ч	C	¥	IJ	V	IJ	S
	Alternative site	U	A	C	A	V	V	G	C	F	IJ	A	G	U	H	G	C	U
	Genome	r .	(7)	(5)	0	0	(7)	-	f .	(7	~	r.,	0	-	(7)	_	(5)	0
	ysical I ittion 5p)	23195 T	23222 G	23262 C	23268 C	15356 C	15498 C	45513 A	45530 T	46684 C	46808 A	46850 T	16856 C	16949 A	17652 C	56827 A	57664 C	58121 C
ned	me Phr pos (1	282	282	282	282	284	284	284	284	284	284	284	284	284	284	28¢	286	286
Continu	Chromoso	0 Chr12																
TABLE 2	Gene ID	AH12G0237	AH12G0237	AH12G0237	AH12G0237	AH12G0239	AH12G0241	AH12G0241	AH12G0241									

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	ne function	ve disease nce RPP13- otein 1	cyclopropane- oxylate se-like 1	don-like n B1	ve disease nce RPP13- otein 1	membrane ors 3BATP g	membrane ors 3BATP g	membrane ors 3BATP g	n of unknown m (DUF594)	n of unknown m (DUF595)	ve disease nce RPP13- otein 1	family ular one regulator	family ular °one regulator	family ular °one regulator	/threonine- n phosphatase	otransferase- C plant mobile n family 1	R family osase	
	Gei	Putatr resista like pr	1- amino 1-carb syntha proteii	Reticu proteii	Putati ^r resista like pr	Transı recept bindin	Transı recept bindin	Transı recept bindin	Proteii functio	Proteii functio	putativ resista like pr	BAG f molec chaper 6	BAG f moleci chaper 6	BAG f molecr chaper 6	Serine proteii 7	Amine like 20 domai protei	MuDF transp	
	L99 (99% confidence interval lower side)	-0.584825	-0.584916	-0.585057	-0.58527	-0.585315	-0.585315	-0.585315	-0.585331	-0.585331	-0.585359	-0.585372	-0.585372	-0.585372	-0.585391	-0.585391	-0.585158	
	U99 (99% confidence interval upper side)	0.582471	0.582557	0.582693	0.582914	0.582977	0.582977	0.582977	0.583012	0.583012	0.583117	0.583131	0.583131	0.583131	0.583152	0.583152	0.582958	
	L95 (95% confidence interval lower side)	-0.457961	-0.458031	-0.458138	-0.458315	-0.458361	-0.458361	-0.458361	-0.458386	-0.458386	-0.4584	-0.458403	-0.458403	-0.458403	-0.458406	-0.458406	-0.45814	
	U95 (95% confidence interval upper side)	0.456704	0.456781	0.456889	0.457055	0.457096	0.457096	0.457096	0.457119	0.457119	0.457106	0.457108	0.457108	0.457108	0.457108	0.457108	0.456817	
	acid hange	aa/ aa	ga/ ga	g/tCg	<u>ð</u> 0	gt/Agt	T/ttA	at/Gat	tt/Gtt	tg/Atg	Ct/aGt	at/Cat	gt/Cgt	T/atG	3t/cAt	Ga/ Aa	gt/ gt	
	ion effect A	GG	DUS_CODING G	DUS_CODING 17	DUS_CODING tg	OUS_CODING C	DUS_CODING II	OUS_CODING C	OUS_CODING CI	OUS_CODING CI	DUS_CODING aC	OUS_CODING G	DUS_CODING T _i	DUS_CODING at	OUS_CODING 60	c/ 002_CODING cC	DUS_CODING G	
	SNP substitut	DMYNONY2_NON	omynony2, non	NON_SYNONYMC)WANONAS [_] NON	NON_SYNONYMC	NON_SYNONYMC	NON_SYNONYMC	NON_SYNONYMC	NON_SYNONYMC	NON_SYNONYMC	JWANONY2_NON	NON_SYNONYMC	NON_SYNONYMC	DWANONAS ⁻ NON	JWANONY2_NON	NON_SNONYMC	
	Delta SNP- index (RB SNP- index-SB SNP- index)	0.64	0.71	0.64	6.79	0.43	0.43	06.0	0.78	0.73	16:0	0.82	0.73	0.80	0.70	0.58	0.64	
	SNP- index of SB	0.18	0.17	0.17	0.14	0.57	0.50	0.10	0.22	0.22	0.09	0.18	0.27	0.20	0.15	0.20	0.36	
	SB Depths of Ref, Alt	3,14	3,15	2,10	3,18	4,3	4,4	1,9	2,7	2,7	1,10	4,18	4,11	1,4	2,11	4,16	5,9	
	Number of reads covering the site (X cov- erage) in sus- ceptible bulk (SB)	17	18	12	21	~	∞	10	6	6	Ξ	22	15	ى ئ	13	20	14	
	Susceptible bulk (SB) base	s	X	Y	S	М	M	S	s	М	S	s	X	К	R	Y	Я	
	SNP- index of RB	0.81	0.88	0.81	0.94	1.00	0.93	1.00	1.00	0.95	1	1.00	1.00	1.00	0.86	0.78	1.00	
	RB Depths of Ref, Alt	22,5	21,3	21,5	15,1	17,0	25,2	5,0	24,0	21,1	13,0	13,0	17,0	18,0	18,3	14,4	20,0	
	Number of reads covering the site (X cov- erage) in resis- tant bulk (RB)	27	24	26	16	17	27	ы	24	22	13	13	17	18	21	18	20	
	Resistant bulk (RB) base	s	Y	Y	IJ	U	н	U	O	O	U	U	Т	Т	К	Y	U	
	Alternative site	U	F	U	C	¥	¥	IJ	IJ	V	IJ	U	U	U	×	H	A	
	Genome																	
	p)	3129 C	2570 C	7519 T	5550 G	7238 C	7384 T	3195 C	2506 C	2515 C	9638 C	5337 G	5406 T	5462 T	691 G	3516 C	8818 G	
ed	te Phy: positi (bj	2868	2952	3067	3276	3347	3347	3348	3392	3392	3693	3715	3715	3715	3745	3746	3996	
Continu	Chromoson	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	
TABLE 2	Gene ID	AH12G02410	AH12G02520	AH12G02650	AH12G02880	AH12G02960	AH12G02960	AH12G02960	AH12G03000	AH12G03000	AH12G03230	AH12G03240	AH12G03240	AH12G03240	AH12G03290	AH12G03300	AH12G03500	

TABLE 2	Continue	q																		
Gene ID	Chromosome	Physical position (bp)	Reference Genome	Alternative site	Resistant bulk (RB) base	Number of reads covering 1 the site (X cov- e erage) in resis- tant bulk (RB)	RB 5 Depths i of Ref, o Alt	sNP- ndex f RB	bulk (SB) r bulk (SB) r base th	Number of ads covering I te site (X cov- rage) in sus- ceptible bulk (SB)	SB Depths of Ref, Alt	SNP- I index i	belta SNP- ndex (RB SNP- index-SB SNP- SNP- index)	SNP substitution effect	Amino acid change	U95 (95% confidence interval upper side)	L95 (95% confidence interval lower side)	U99 (99% confidence interval upper side)	L99 (99% confidence interval lower side)	Gene function
AH12G03500	Chr12	3999460	U	L	К	25	22,3	0.88 F		17	4,13	0.24	0.64	NON_SYNONYMOUS_CODING	aGg/ aTg	0.456817	-0.45814	0.582958	-0.585158	MuDR family ransposase
AH12G03500	Chr12	3999687	A	IJ	R	22	20,2	0.91 0	(7)	17	0,17	0.00	0.91	NON_SYNONYMOUS_CODING	Aac/ Gac	0.46	-0.46	0.58	-0.59	MuDR family ransposase
AH12G03560	Chr12	4096307	A	F	¥	15	15,0	L 00.1		ιń	0,5	0.00	1.00	NON_SYNONYMOUS_CODING	gTg/ gAg	0.46	-0.46	0.58	-0.59	UDP- glycosyltransferase 72B1
AH12G03600	Chr12	4142557	Т	IJ	K	17	15,2	1 88.0		7	1,6	0.14	0.74	NON_SYNONYMOUS_CODING	Tct/Gct	0.46	-0.46	0.58	-0.59	Disease resistance protein TAO1
AH12G03600	Chr12	4142566	IJ	V	R	17	15,2	98.0	_	5	0,5	0.00	0.88	NON_SYNONYMOUS_CODING	Gtt/Att	0.46	-0.46	0.58	-0.59	Disease resistance protein TAO1
AH12G03600	Chr12	4142572	IJ	V	R	17	15,2	0.88	_	5	0,5	0.00	0.88	NON_SYNONYMOUS_CODING	Ggt/ Agt	0.46	-0.46	0.58	-0.59	Disease resistan <i>ce</i> protein TAO1
AH12G03900	Chr12	4633134	O	Т	Y	19	17,2	680	N.	15	2,13	0.13	0.76	Non_sy nony mous_coding	Gcg/ Acg	0.46	-0.46	0.58	-0.58	Oytochrome b561 und DOMON domain-containing protein
AH12G03980	Chr12	4691788	C	Т	Υ	20	16,4	0.8	ν.	17	4,13	0.24	0.56	STOP_GAINED	Caa/ Taa	0.456446	-0.457759	0.582353	-0.585012	nomolog of histone chaperone HIRA
AH12G03980	Chr12	4696605	IJ	V	IJ	25	25,0	1.00 F	~	12	2,10	0.17	0.83	NON_SYNONYMOUS_CODING	Gtt/Att	0.456446	-0.457759	0.582353	-0.585012	nomolog of histone chaperone HIRA
AH12G04010	Chr12	4718691	U	L	К	21	16,5	0.76 1	r.,	16	1,15	0.06	0.70	Non_synonymous_coding	Ctg/Atg	0.456475	-0.457785	0.582373	-0.585056	Eukaryotic aspartyl protease family protein
AH12G04170	Chr12	5007281	Т	¥	H	23	23,0	1.00	~	22	19,3	0.86	0.14	Non_sy nony mous_coding	cAc/ cTc	0.457503	-0.45874	0.58321	-0.586473	Endosomal argeting BRO1-like domain-containing protein
AH12G04170	Chr12	5007411	¥	O	×	26	26,0	1.00	V	22	17,5	0.77	0.23	Non_sy nony mous_coding	Tca/ Gca	0.457503	-0.45874	0.58321	-0.586473	Endosomal argeting BRO1-like domain-containing protein
AH12G04170	Chr12	5007427	U	O	U	27	27,0	1.00 S		22	17,5	0.77	0.23	Non_sy nony mous_coding	ttC/ttG	0.457503	-0.45874	0.58321	-0.586473	Endosomal argeting BRO1-like domain-containing protein
AH12G04170	Chr12	5007448	C	U	O	24	24,0	1.00 S		18	15,3	0.83	0.17	Non_sy nony mous_coding	tgG/ tgC	0.457503	-0.45874	0.58321	-0.586473	3ndosomal argeting BRO1-like domain-containing protein
AH12G04170	Chr12	5007454	O	U	U	27	27,0	1.00 S		17	14,3	0.82	0.18	Non_sy nony mous_coding	ttG/ttC	0.457503	-0.45874	0.58321	-0.586473	Endosomal argeting BRO1-like domain-containing protein
AH12G04330	Chr12	5187080	U	Ł	O	20	18,2	060	N.	10	3,7	0.30	0.60	Non_synony mous_coding	Gag/ Aag	0.458456	-0.459622	0.584098	-0.587787	2-Joop containing nucleoside riphosphate ŋydrolases uperfamily protein
AH12G04490	Chr12	55 262 42	U	Т	Y	25	19,6	0.76		14	4,10	0.29	0.47	Non_sy nony mous_coding	gGa/ gAa	0.460031	-0.461049	0.585491	-0.5901	Uncharacterized protein

	Gene function	aden osyl-L- ethionine- pendent ethyltransferases perfamily protein	ncharacterized otein	ncharacterized otein	ketoacyl-CoA nthase 1	ncharacterized otein	ruvate hydrogenase E1 mponent subunit bha-3/2C loroplastic	obably inactive acine-rich repeat ceptor-like otein kinase	obably inactive ucine-rich repeat ceptor-like otein kinase	obably inactive acine-rich repeat ceptor-like otein kinase	obably inactive acine-rich repeat zeptor-like otein kinase	obably inactive acine-rich repeat ceptor-like otein kinase	avone 3'-O- ethyltransferase 1	avone 3'-O- ethyltransferase 1	avone 3'-O- șthyltransferase 1	avone 3'-O- ethyltransferase 1	t domain- ntaining RNAion factor 3A1
	L99 (99% onfidence interval wer side)	.0.591388 S- m de suy	-0.591789 U.	0.592101 U.	-0.59291 3 syr	0.594617 U.	0.594957 P) dé co alţ ch	-0.595523 P1 let ret pr	-0.595523 P1 let ret	-0.595523 P1 let ret	-0.595523 P1 let ret	-0.595747 P1 let ret	-0.595763 Fl. m	-0.595763 Fl.	-0.595763 Fl.	-0.595823 Fl.	-0.59586 B:
	U99 (99% I onfidence o interval pper side) lo	0.58614 -	0.586301 -	0.586503 -	0.587129	0.588336	0.588597	0.589041	0.589041	0.589041	0.589041	0.589192	0.589199	0.589199	0.589199	0.58922	0.589187
	L95 (95% confidence e interval lower side) u	-0.461707	-0.461893	-0.462076	-0.462593	-0.46355	-0.463716	-0.463988	-0.463988	-0.463988	-0.463988	-0.464051	-0.46405	-0.46405	-0.46405	-0.464034	-0.463869
	U95 (95% confidence interval upper side)	0.460793	0.46102	0.461224	0.461787	0.462862	0.463056	0.463369	0.463369	0.463369	0.463369	0.463457	0.463458	0.463458	0.463458	0.463453	0.463323
	Amino acid change	aaC/ aaA	Ggt/Tgt	Gtg/ Atg	Tcc/ Ccc	cCc/ cTc	cAg/ cGg	Ttc/Ctc	gAc/ gTc	tCt/tAt	ttT/ttA	gCg/ gTg	aCa/ aTa	Atc/Ttc	aCt/aGt	gCt/gTt	Gec
	SNP substitution effect	NON_SY NONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING
	Delta SNP- index (RB SNP- index-SB SNP- index)	0.78	0.60	0.87	0.40	0.56	0.62	0.11	0.17	0.17	0.29	0.62	0.77	0.23	0.21	0.57	0.93
	SNP- index of SB	0.17	0.21	0.08	0.31	0.21	0.17	0.89	0.83	0.83	0.71	0.14	0.00	0.77	0.71	0.30	0.00
	SB Depths of Ref, Alt	1,5	3,11	1,12	5,11	3,11	3,15	8,1	10,2	10,2	5,2	3,18	0,13	10,3	10,4	3,7	0,9
	Number of reads covering the site (X cov- erage) in sus- ceptible bulk (SB)	Q	14	13	16	14	18	6	12	12	м	21	13	13	14	10	6
	Susceptible bulk (SB) base	К	К	R	Υ	Υ	К	Υ	×	М	M	ж	¥	M	S	R	U
	SNP- index of RB	0.94	0.81	0.94	0.71	0.77	620	1.00	1.00	1.00	1.00	0.76	0.77	1.00	0.92	0.87	0.93
	RB Depths of Ref, Alt	17,1	17,4	17,1	17,7	17,5	15,4	12,0	14,0	15,0	18,0	13,4	17,5	20,0	23,2	13,2	13,1
	Number of reads covering the site (X cov- erage) in resis- tant bulk (RB)	18	21	18	24	22	19	12	14	15	18	17	22	20	25	15	14
	Resistant bulk (RB) base	U	К	IJ	Υ	Υ	2	L	V	U	T	К	В	Т	IJ	R	К
	Alternative site	Ŧ	Г	V	U	Г	U	O	H	V	ĸ	V	A	¥	U	V	U
	Reference Genome	U	U	c	н	U	<	H	A	0	н	U	G	ц	G	G	н
	hysical osition (bp)	765500	364398 (11414 (18424	\$21518 (391830	17918	517937	517943	518088	594841	505230 (506476	507843 ((55924 (343052
nued	pd pd	57	58	59	60	63	63	65	65	65	65	65	99	99	99	99	68
Contin	Chromos	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12	Chr12
TABLE 2	Gene ID	AH12G04580	AH12G04670	AH12G04700	AH12G04740	AH12G04980	AH12G05040	AH12G05080	AH12G05080	AH12G05080	AH12G05080	AH12G05100	AH12G05120	AH12G05120	AH12G05120	AH12G05140	AH12G05250

TABLE 2	Continue	Q																		
Gene ID	Chromosome	Physical position (bp)	Reference Genome	Alternative site	Resistant bulk (RB) base	Number of reads covering the site (X cov- erage) in resis- tant bulk (RB)	RB Depths of Ref, Alt	SNP- index of RB	Susceptible bulk (SB) base	Number of reads covering the site (X cov- erage) in sus- ceptible bulk (SB)	SB Depths of Ref, Alt	of SB	Delta SNP- index (RB SNP- index-SB SNP- index)	SNP substitution effect	Amino acid change	U95 (95% confidence interval upper side)	L95 (95% confidence interval lower side)	U99 (99% confidence interval upper side)	L99 (99% confidence interval lower side)	Gene function
AH12G05320	Chr12	6989102	¥	9	R	10	1,6	060	R	12	2,10	0.17	0.73	on_synonymous_coding	Tcc/ Ccc	0.462905	-0.463448	0.588882	-0.5955	Serine/threonine- protein phosphatase 7
AH12G05420	Chr12	7082034	H	C	F	16	16,0	1.00	Υ	15	1,14	0.07	0.93	NON_SYNONYMOUS_CODING	Aca/ Gca	0.462547	-0.463075	0.588567	-0.595158	Probable inactive purple acid phosphatase 29
AH12G05820	Chr12	7671331	U	IJ	C	23	22,1	0.96	s	7	1,6	0.14	0.81	NON_SYNONYMOUS_CODING	caC/ caG	0.460651	-0.461098	0.586887	-0.593277	FRIGIDA-like protein 3
AH12G05970	Chr12	7992939	IJ	V	К	12	11,1	0.92	ж	15	2,13	0.13	0.78	Non_SynonyMous_coding	Cgc/ Tgc	0.459751	-0.460171	0.585875	-0.592182	Aminotransferase- like 2C plant mobile domain family protein
AH12G06150	Chr12	8202579	U	F	H	21	0,21	1.00	X	17	13,4	0.24	0.76	NON_SYNONYMOUS_CODING	Gaa/ Aaa	0.459288	-0.459699	0.585408	-0.591564	AT-rich interactive domain-containing protein 4
AH12G06180	Chr12	8256014	C	V	C	21	20,1	0.95	М	21	3,18	0.14	0.81	NON_SYNONYMOUS_CODING	Cat/Aat	0.45923	-0.45964	0.585372	-0.591479	RING/U-box superfamily protein
AH12G06180	Chr12	8260084	IJ	V	Ð	22	22,0	1.00	R	6	6,3	0.67	0.33	NON_SYNONYMOUS_CODING	aGg/ aAg	0.459218	-0.459627	0.585364	-0.591462	RING/U-box superfamily protein
AH12G06180	Chr12	8260159	F	U	Т	16	16,0	1.00	Υ	14	4,10	0.29	0.71	NON_SYNONYMOUS_CODING	tTt/tCt	0.459218	-0.459627	0.585364	-0.591462	RING/U-box superfamily protein
AH12G06180	Chr12	8260203	IJ	V	IJ	18	18,0	1.00	R	24	15,9	0.63	0.38	NON_SYNONYMOUS_CODING	Gtt/Att	0.459218	-0.459627	0.585364	-0.591462	RING/U-box superfamily protein
AH12G06180	Chr12	8271833	A	IJ	A	20	19,1	0.95	R	18	6,12	0.33	0.62	NON_SYNONYMOUS_CODING	tAt/tGt	0.459205	-0.459615	0.585355	-0.591445	RING/U-box superfamily protein
AH12G06180	Chr12	8271850	IJ	O	ŋ	22	21,1	0.95	S	18	6,12	0.33	0.62	NON_SYNONYMOUS_CODING	Gca/ Cca	0.459205	-0.459615	0.585355	-0.591445	RING/U-box superfamily protein
AH12G06210	Chr12	8303749	A	IJ	A	26	26,0	1.00	R	9	5,1	0.83	0.17	NON_SYNONYMOUS_CODING	Aca/ Gca	0.459187	-0.459596	0.585344	-0.591416	RING/U-box superfamily protein
AH12G06210	Chr12	8303776	A	IJ	A	23	23,0	1.00	R	9	3,3	0.50	0.50	NON_SYNONYMOUS_CODING	Aaa/ Gaa	0.459187	-0.459596	0.585344	-0.591416	RING/U-box superfamily protein
AH12G06250	Chr12	8416831	F	IJ	Т	16	16,0	1.00	К	9	1,5	0.17	0.83	non_synonymous_coding	Tgt/Ggt	0.459105	-0.459511	0.585247	-0.591281	RING/U-box superfamily protein
AH12G06300	Chr12	8578649	C	Т	Υ	16	13,3	0.81	Y	16	3,13	0.19	0.63	non_synonymous_coding	gCt/gTt	0.458995	-0.459395	0.585101	-0.591125	Exocyst subunit exo70 family protein B2
AH12G06300	Chr12	8579347	U	Н	C	12	12,0	1.00	Y	18	3,15	0.17	0.83	Non_synonymous_coding	Cgg/ Tgg	0.458995	-0.459395	0.585101	-0.591125	Exocyst subunit exo70 family protein B2
AH12G06320	Chr12	8636540	IJ	F	IJ	18	18,0	1.00	н	6	6'0	0.00	1.00	NON_SYNONYMOUS_CODING	aaC/ aaA	0.458954	-0.459351	0.585052	-0.591055	Putative disease resistance RPP13- like protein 1
AH12G06320	Chr12	8636563	C	V	C	18	18,0	1.00	V	œ	0,8	0.00	1.00	NON_SYNONYMOUS_CODING	Gat/Tat	0.458954	-0.459351	0.585052	-0.591055	Putative disease resistance RPP13- like protein 1
AH12G06320	Chr12	8636575	U	¥	U	13	13,0	1.00	۲	~	0,7	00.00	1.00	Non_sy Nony Mous_coding	Cac/ Tac	0.458954	-0.459351	0.585052	-0.591055	Putative disease resistance RPP13- like protein 1

	Gene function	Duplicated homeodomain-like superfamily protein	Endosomal targeting BRO1-like domain-containing protein	GDSL esterase/ lipase 5	GDSL esterase/ lipase 5	GDSL esterase/ lipase 3	UDP- glycosyltransferase 72B3	Pentatricopeptide repeat-containing protein	Pentatricopeptide repeat-containing protein	Squamosa promoter-binding- like protein 8	mRNAion regulators
	L99 (99% confidence interval lower side)	-0.590896	-0.590847	-0.5908	-0.590776	-0.590701	-0.590474	-0.589328	-0.589328	-0.589016	-0.588646
	U99 (99% confidence interval upper side)	0.584957	0.584923	0.58489	0.584873	0.584819	0.584691	0.584069	0.584069	0.583973	0.58382
	1.95 (95% confidence interval lower side)	-0.459253	-0.459221	-0.45919	-0.459174	-0.459123	-0.458987	-0.458489	-0.458489	-0.458444	-0.458357
	U95 (95% confidence interval upper side)	0.45886	0.458829	0.458799	0.458783	0.458733	0.458603	0.458109	0.458109	0.458059	0.457966
	Amino acid change	Cta/Ata	aGa/ aAa	aCa/ aAa	caG/ caC	Tta/Gta	gAt/gCt	tGg/ tCg	aAt/aGt	tGc/tAc	Gca/ Tca
	SNP substitution effect	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING	NON_SYNONYMOUS_CODING
	Delta SNP- index (RB SNP- index-SB SNP- index)	0.76	0.82	0.49	0.78	0.46	0.78	0.28	0.33	0.71	0.60
	SNP- index of SB	0.16	0.18	0.44	0.22	0.54	0.12	0.67	0.61	0.17	0.28
	SB Depths of Ref, Alt	3,16	2,9	4,5	2,7	7,6	23,3	8,4	11,7	5,1	8,21
	Number of reads covering the site (X cov- erage) in sus- ceptible bulk (SB)	19	11	6	6	13	26	12	18	Q	29
	Susceptible bulk (SB) base	М	Υ	K	s	K	К	S	Y	К	Ж
	SNP- index of RB	0.92	1.00	0.94	1.00	1.00	060	0.95	0.94	0.88	0.88
	RB Depths of Ref, Alt	11,1	14,0	15,1	18,0	21,0	2,18	19,1	15,1	1.7	21,3
	Number of reads covering the site (X cov- erage) in resis- tant bulk (RB)	12	14	16	18	21	20	20	16	œ	24
	bulk (RB) base	М	O	IJ	C	Т	U	C	Н	В	Ж
	Alternative site	¥	H	Н	IJ	IJ	IJ	IJ	U	V	F
	Reference Genome	U	U	U	C	F	Н	C	F	U	U
	hysical osition (bp)	714663	735310	758727	767570	792582	898415	324156	324470	478950	663318
2 Continued	Chromosome F P	30 Chr12 8	50 Chr12 8	80 Chr12 8	80 Chr12 8	90 Chr12 8	60 Chr12 8	50 Chr12 9	50 Chr12 9	20 Chr12 9	90 Chr12 9
TABLE	Gene ID	AH12G064	AH12G064	AH12G064	AH12G064	AH12G064	AH12G065	AH12G069	AH12G069	AH12G070	AH12G070

TABLE 3 Identification of InDels in putative candidate genes in the genomic region for resistance to on chromosome 12.

Gene	Chromosome	Physical position (bp)	Reference Genome	Alternative site	Resistant bulk (RB) base	Number of reads covering the site (X coverage) in resis- tant bulk (RB)	RB Depths of Ref, Alt	SNP- index of RB	Susceptible bulk(SB)base	Number of reads cov- ering the site (X cov- erage) in susceptible bulk(SB)	SB Depths of Ref, Alt	SNP- index of SB	delta SNP- index (RB SNP- index- SB SNP- index)	SNP substitution effect	U95 (95% confidence interval upper side)	L95 (95% confidence interval lower side)	U99 (99% confidence interval upper side)	L99 (99% confidence interval lower side)	Gene Function
AH12G015	0 Chr12	1888286	TCC	Т	тсс	23	22,1	0.96	TCC,T	9	7,2	0.78	0.18	FRAME_SHIFT	0.499429	-0.498581	0.630045	-0.613463	putative disease resistance protein At3g14460 isoform X2 [Arachis ipaensis]
AH12G016	0 Chr12	2019127	CCTATTCTACG CCTCACAAAATCT	С	CCTATTCTACG CCTCACAAAATCT	28	26,2	0.93	CCTATTCTACGCCTCACAAAATCT, C	22	17,5	0.77	0.16	FRAME_SHIFT	0.499034	-0.498174	0.629563	-0.613179	HXXXD-type acyl-transferase family protein
AH12G016	0 Chr12	2020084	Т	TGGTGAAACA	Т	21	21,0	1.00	T,TGGTGAAACA	15	10,5	0.67	0.33	CODON_INSERTION	0.498917	-0.498054	0.629416	-0.613083	HXXXD-type acyl-transferase family protein
AH12G0192	0 Chr12	2473096	А	ACGTTT	A,ACGTTT	22	19,3	0.86	A,ACGTTT	16	2,14	0.13	0.74	FRAME_SHIFT	0.496175	-0.495264	0.626054	-0.611224	Putative disease resistance protein At3g14460
AH12G0198	0 Chr12	2534345	CTCCGTACCAAG	С	C,CTCCGTACCAAG	24	22,2	0.92	C,CTCCGTACCAAG	20	18,2	0.90	0.02	FRAME_SHIFT	0.495902	-0.494987	0.625693	-0.611031	Putative disease resistance RPP13-like protein 1
AH12G0198	0 Chr12	2539604	ТА	т	ТА	24	24,0	1.00	TA,T	20	17,3	0.85	0.15	FRAME_SHIFT	0.495902	-0.494987	0.625693	-0.611031	Putative disease resistance RPP13-like protein 1
AH12G0198	0 Chr12	2539716	С	CCA	С	25	25,0	1.00	CCA,C	25	15,10	0.60	0.40	FRAME_SHIFT	0.495902	-0.494987	0.625693	-0.611031	Putative disease resistance RPP13-like protein 1
AH12G0209	0 Chr12	2620516	TGG	Т	TGG	24	24,0	1.00	T,TGG	9	8,1	0.89	0.11	FRAME_SHIFT	0.495698	-0.494779	0.625463	-0.610928	Putative disease resistance RPP13-like protein 1
AH12G0239	0 Chr12	2847658	Т	ТСААА	Т	16	15,1	0.94	T,TCAAA	6	2,4	0.33	0.60	FRAME_SHIFT	0.495495	-0.494571	0.625259	-0.610874	Putative disease resistance RPP13-like protein 1
AH12G0244	0 Chr12	2879448	Т	TGAG	Т	11	11,0	1.00	T,TGAG	6	5,1	0.83	0.17	CODON_CHANGE_PLUS_CODON_INSERTION	0.495507	-0.494586	0.625287	-0.610914	Putative disease resistance RPP13-like protein 1
AH12G0260	0 Chr12	3030085	Т	TGATGGTGAGA CTTTGAAGAGCA ACCAGTCCTTC TCTTGGAAATGAC ACAAATGAGTTGCAGTGATG	TGATGGTGAGACTTT GAAGAGCAACCAGTC CTTCTCTTGGAAATGA CACAAATGAGTTGCAGTGATG, T	8	7,1	0.88	T,TGATGGTGAG ACTTTGAAGAGCA ACCAGTCCTTCTC TTGGAAATGACAC AAATGAGTTGCAGTGATG	13	2,11	0.15	0.72	CODON_INSERTION	0.495575	-0.494661	0.625474	-0.61112	Putative disease resistance RPP13-like protein 1
AH12G0324	0 Chr12	3715392	ATCG	A	ATCG	17	17,0	1.00	A,ATCG	17	4,13	0.24	0.76	CODON_DELETION	0.496363	-0.495453	0.627006	-0.612884	BAG family molecular chaperone regulator 6
AH12G061	0 Chr12	8165877	G	GGATTAGTGGTGCAGCAGCTTGT	G	20	20,0	1.00	G, GGATTAGTGGTGCAGCAGCTTGT	10	3,7	0.30	0.70	FRAME_SHIFT	0.502752	-0.502787	0.64146	-0.630466	hypothetical protein MTR_5g086360
AH12G0618	0 Chr12	8289562	СТАА	С	CTAA	10	10,0	1.00	CTAA,C	21	19,2	0.90	0.10	CODON_DELETION	0.502481	-0.502506	0.641028	-0.630118	E3 ubiquitin- protein ligase RNF144A-like



seq approach is an increasingly popular sequencing-based method for the identification of candidate genomic regions associated with target traits in peanut (Varshney et al., 2019; Pandey et al., 2020). As it only requires whole genome sequences of parents and extreme trait bulks from the mapping population, it is economical, efficient, fast, and cost-effective (Takagi et al., 2013). Traditional QTL mapping methods are limited in the fine mapping of target genes and QTLs because they lack both high-



FIGURE 5

Validation of putative candidate gene-based markers of bacterial wilt resistance. (A) The SNP marker validation of candidate gene *AH12G03230* using a validation set comprising the resistant parent YY92, susceptible parent XHXL, and susceptible and resistant RIL lines). (B) SNP variation in the *AH12G03230* gene. (C) The *AH12G03230* gene is predicted to encode the CC-NBS-LRR resistance protein. (D) Putative genomic region on Chromosome 12 of *Arachis hypogaea* that encodes resistance to *Ralstonia solanacearum* infections (E) The *AH12G03220* gene is predicted to encode the NBS-LRR resistant protein (E1 to E5 refer to exon numbers while 11 to 14 refer to intron numbers), (F) SNP variation in the *AH12G0600* gene and (G) marker validation on a validation set comprising resistant parent YY92, susceptible parent XHXL, susceptible RIL lines (YX32, YX68, YX160, YX211, YX293, YX554, YX622, YX840 and YX178, R123, R592, YX57, YX80, YX59, YX299, YX469, YX707 and YX939), and resistant RIL lines (YX131, YX189, YX284, YX303, YX636, YX712, YX759, YX905, YX962, YX540, YX544, YX793, YX802, YX875, R160, R201, R215 and R739).

density genetic maps and a series of near-isogenic lines (Pandey et al., 2017). Despite not having a large segregation population as a prerequisite, the QTL-seq approach was successful in identifying candidate genes for many crop traits (Das et al., 2014; Lu et al., 2014; Illa et al., 2015; Singh et al., 2016; Wang et al., 2016; Pandey et al., 2017; Srivastava et al., 2017; Chen et al., 2018; Clevenger et al., 2018; Luo et al., 2019; Bommisetty et al., 2020; Kumar et al., 2020; Lei et al., 2020; Tudor et al., 2020; Zhao et al., 2020; Cao et al., 2021; Dong et al., 2021; Topcu et al., 2021; Wang et al., 2021; Yang et al., 2021; Zhang et al., 2021). In the present study, a QTL-seq approach was successfully applied to identify genomic regions and candidate genes for RRS using resequencing data of both parental genotypes and pooled samples of the RIL population (Yueyou 92×Xinhuixiaoli) (Supplementary Figure 1), which corroborated our previously reported QTL mapping findings (Zhao et al., 2016).

The use of a common reference genome that is associated with deep sequencing and large bulks should result in highly accurate maps. As per the original QTL-seq approach (Takagi et al., 2013), the genome assemblies of either one or both parents were used as reference to analyze the SNP variants in the two extreme bulks based on diploid reference genomes due to the unavailability of the cultivated peanut genome (Pandey et al., 2017; Luo et al., 2019; Kumar et al., 2020; Chen et al., 2021). Candidate genomic regions were then associated with target traits by the Δ SNP index method (Takagi et al., 2013). The choice of the parental reference genome possibly affects this association of candidate genomic regions due to differing levels of alignment errors (Luo et al., 2019). Moreover, the algorithm uses the reference genomes to replace the parental genomes in bigger diversity areas, which may cause erroneous assemblies of the parent genomes, especially for wild diploid ancestors (unpublished data). To increase the reliability of identified genomic regions and candidate genes, we used the Arachis hypogaea Shitouqi genome as a reference. Shitouqi (A. h. fastigiata var. vulgaris), belongs to the subsp. fastigiata, as does the parents of the population, and its high-quality genome sequence was recently reported (Zhuang et al., 2019). Based on a large RIL population of 581 individual lines from the cross of resistant YY92 and susceptible Xinhuixiaoli, 30 resistant and 30 susceptible lines were selected to respectively construct the extreme R and S-bulks (Figure 1 and Supplementary Table 1). We generated 108.38 and 96.92 Gb of sequence reads at a sequencing depth of $38 \times$ and $34 \times$ for the R- and S-bulks, respectively (Table 1). Nearly 98% of these reads were correctly mapped onto the reference genome. High densities of homozygous SNPs (381,642) and InDels (98,918) between parents were identified by resequencing for RRS mapping (Supplementary Tables 2, 3). The candidate region of 5.73 Mb on Chr12 was identified by combining the ED and Δ SNP/ InDel index algorithms for both SNPs and InDels (Figures 2, 5; Table 2; Supplementary Tables 2, 3) at a P<0.01 confidence level. These aided the precise and accurate discovery of candidate genomic regions, genes, and SNPs/Indels markers associated with RRS.

The clear extreme phenotypic differences between parents as well as those of the pooled population were the crucial prerequisite for candidate gene mapping via the QTL-seq approach (Zhao et al., 2020; Chen et al., 2021). An R2R3-MYB transcription factor gene named AhTc1 was mapped and characterized as associated with purple testa via the QTL-seq approach. Allele-specific markers were developed, which demonstrated that the marker pTesta1089 was closely linked with purple testa (Zhao et al., 2020). Chen et al. identified AhRt1 bHLH transcriptional factor as the candidate gene that regulates the red testa color of peanut via QTL-sequencing analyses. An AhRt1 diagnostic marker was then developed for validating and distinguishing different populations and peanut varieties (Chen et al., 2021). The phenotype evaluation of plant disease resistance traits is a challenge for map-based gene cloning. Unlike the testa color phenotype, disease resistance traits are complicated and affected by the environment, especially those for resistance to peanut bacterial wilt. As usual, the identification method in the disease nursery was used for the resistant evaluation. The survival rate was calculated from the number of dead plants until the point of harvesting for QTL mapping of BWR (Luo et al., 2019). However, the natural identification method was affected by the temperature, the soil environment, and anthropogenic effects. In our previous study, artificial inoculation by the leaf-cutting method was successfully used to evaluate the resistance to bacterial wilt via high throughput sequencing for QTL mapping of the cultivated peanut (Zhao et al., 2016). The resistance phenotyping in this study validated the accuracy of our previous findings (Zhao et al., 2016). The disease symptoms were classified into six disease severity ratings, and the resistance level of different lines was calculated by the disease index (Figure 1 and Supplementary Table 1). The phenotype of different lines was truly reflective of the disease resistance as per the DI method and the candidate genomic region was then accurately identified by the QTL-seq approach (Figure 2 and Supplementary Figures 6-9). This method is clearly valuable in the phenotyping of RRS for large populations in a cost-effective and practicable manner.

Hitherto, the main stable QTLs of RRS in the peanut were successfully identified through the original QTL method and QTL-seq approach (Luo et al., 2020). We previously reported SSR and SNP marker-based genetic linkage maps obtained through the classical QTL mapping (Zhao et al., 2016). Two major QTLs (qBW-1 and qBW-2) were identified for RRS, which were in the LG1 and LG10 linkage groups based on RAD- and BSA-seq techniques in F2 plants. One QTL linked to two QTL peaks on ChrB02 was identified in an F8 RIL population (Zhao et al., 2016). Luo et al. reported one QTL named qBWRB02.1 that possibly spanned a 5.14 Mb (0.81-5.95 Mb) interval on chromosome B02 based on its flanking SSR markers (Luo et al., 2020). Via the QTL-seq approach, they then identified a 2.07 Mb genomic region on ChrB02 associated with RRS across three environments (Luo et al., 2020). Two adjacent genomic regions (2.81-4.24 Mb and 6.54-8.75 Mb) on chromosome B02 were identified within the confidential interval of qBWRB02-1 and

thus designated as *qBWRB02-1-1* and *qBWRB02-1-2* based on two diploids reference genomes (Luo et al., 2019). In the present study, by using the QTL-seq approach (Supplementary Figure 1), a major peak on Chr12 spanning a 7.2 Mb (1.8-9.0 Mb) interval with a confidence level of P<0.05, 5.73 Mb of this peak had a confidence level of P<0.01 was identified as the candidate genomic region for RRS, corroborating RAD-seq findings (Zhao et al., 2016) and SLAF-seq techniques (Figure 3 and Supplementary Figure 10, unpublished). These revealed the precise identification of the candidate genomic region via the QTL-seq approach. In China, peanut cultivars that are resistant to bacterial wilt, originate from Xiekangqing, Taishan Zhenzhu, and the wildtype species (A.diogoi). Two major QTLs that were both named qBWRB02.1, were identified from the cross of Yuanza 9102 × Xuzhou 68-4. Yuanza 9102 is a popular resistant cultivar whose resistance stemmed from A.diogoi (Luo et al., 2020). Two major QTLs, qBWRB02-1-1 and qBWRB02-1-2, were fine-mapped from the cross of Zhonghua $6 \times$ Xuhua 13. The resistance phenotype of Zhonghua 6 stemmed from the Chinese landrace Taishan Zhenzhu (Luo et al., 2019) whereas that of Yueyou92 stemmed from the Chinese landrace Xiekangqing, which is the parental type for many RRS breeding programs in South China (Janila et al., 2016; Luo et al., 2020). These candidate genomic regions associated with RRS were also mapped onto an interval of 10 Mb on chromosome 12.

QTL-seq approach was demonstrated as an effective method for the identification of putative SNPs associated with RRS. These could be developed into allele markers by using either different genotypes or diagnostic markers after the validation (Luo et al., 2019). Allele-specific markers that can be identified via agarose gel electrophoresis are the most cost-effective assays for genotyping a breeding population to select plants with the desired allele (Pandey et al., 2017). Here, 1807 effective SNPs and 629 InDels were identified. They span a 7.2 Mb genomic region on chromosome 12 that is associated with RRS. A total of 180 nonsynonymous SNPs and 14 InDels respectively affected 75 and 11 candidate genes that encode RRS (Tables 2, 3, Supplementary Tables 4 and 5). The putative RRS-encoding NBS-LRR gene had 44 SNPs, which were targeted for the development of allele-specific markers. Despite designing primers for both alleles of each SNP, amplification of markers was often observed for only one of the allele pairs. Nonamplification of a few markers may be due to DNA template-primer mismatches (You et al., 2008; Pandey et al., 2017). In this study, polymorphic SNP markers were selected as diagnostic markers. Of the 30 amplified markers, two markers (qRRS18 and qRRS19) were robust and co-segregated with RRS (Figure 5). These two polymorphic markers were then validated on a panel of diverse genotypes containing naturally resistant parental types (Yueyou 92) of the RIL population, 18 introgression lines, susceptible parental types (Xinhuixiaoli), and 18 susceptible RIL lines. The 'qRRS18' marker amplified susceptible alleles, whereas

the '*qRRS19*' marker amplified resistant alleles. Thus, these diagnostic allele markers can be applied in MAS for RRS in peanut breeding programs.

In plants, R genes play an important role in defending against pathogens through activation of the innate immune system (Jones and Dangl, 2006; Zhang et al., 2017; Sun et al., 2020; Chang et al., 2022). Most of them belong to the NBS-LRR type, which has been identified in many crops by map-based cloning (Takken and Joosten, 2000; Hulbert et al., 2001; McDowell and Woffenden, 2003; Meyers et al., 2005). The function of the NBS-LRR genes was correlated with either protein length or SNP variants. RRS1-R was the first reported Tir-NBS-LRR gene that conferred resistance to bacterial wilt in Arabidopsis species (Deslandes et al., 2003). RRS1-S, the allele of RRS1-R found in susceptible species, encode a protein without the WRKY domain. The two genes encoded contrasting phenotypes after R. solanacearum infections (Deslandes et al., 1998; Deslandes et al., 2003). Deng et al. identified an NBS-LRR gene named PigmR. It conferred resistance to the fungus Magnaporthe oryzae in rice and its encoded protein lacked four amino acids in the leucine-rich repeat (LRR) domain when compared to the R4 gene that conferred a susceptible phenotype (Deng et al., 2017). In the present study, the predicted products of 22 of the 180 candidate genes with nonsynonymous mutations in the 7.2 Mb region, were NBS-LRR type disease resistance proteins (Table 2 and Figure 4). Notably, eight of the 22 candidate NBS-LRR genes were identified at a high confidence level as associated with RRS (Figure 4). Moreover, seven NBS-LRR genes had SNP variant sites in the LRR domain (Figure 4). The diagnostic SNP markers (gRRS18 and gRRS19) of the candidate AH12G03230 and AH12G06320 NBS-LRR genes were validated in the RIL lines (Figure 5). This indicates that AH12G03230 and AH12G06320 might be the candidate resistant genes for RRS in cultivated peanut. Therefore, based on these findings, these putative resistance genes possibly significantly contribute to RRS in peanut and should thus be targeted as candidates for fine mapping and function validation.

5 Conclusion

In this study, the QTL-seq approach was proven as a powerful method for the successful identification of genomic regions and candidate genes of major and robust QTLs that are associated with RRS. We not only identified a 7.2 Mb genomic region on chromosome 12 containing eight candidate NBS-LRR resistance genes but also availed validated allele-specific diagnostic markers and key candidate genes for RRS breeding. The genomic information (genes) and tools (markers) could be used in genomics-assisted breeding programs to accelerate the development of peanut varieties with enhanced RRS as well as to increase insights into RRS molecular mechanisms.

Data availability statement

The original contributions presented in the study are publicly available. This data can be found here: NCBI, PRJNA851221.

Author contributions

WZ and RV conceived the original research plan and designed the experiment. CZ, WX, HF, YC, HC, TC, QY, YZ, KC, and XZ performed the experiments and analyzed the data. CZ and WX wrote the manuscript, while WZ, MP, and RV reviewed and edit the manuscript. All authors analyzed the data and approved the submitted version.

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Conflict of interest

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

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Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/ fpls.2022.1048168/full#supplementary-material

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