



# A Genome-Wide Association Study of Coleoptile Length in Different Chinese Wheat Landraces

Jun Ma<sup>1†</sup>, Yu Lin<sup>2†</sup>, Si Tang<sup>2</sup>, Shuonan Duan<sup>1</sup>, Qing Wang<sup>2</sup>, Fangkun Wu<sup>2</sup>, Caixia Li<sup>2</sup>, Xiaojun Jiang<sup>2</sup>, Kunyu Zhou<sup>2</sup> and Yaxi Liu<sup>2,3\*</sup>

<sup>1</sup> College of Agronomy and Biotechnology, China Agricultural University, Beijing, China, <sup>2</sup> Triticeae Research Institute, Sichuan Agricultural University, Chengdu, China, <sup>3</sup> State Key Laboratory of Crop Gene Exploration and Utilization in Southwest China, Chengdu, China

#### **OPEN ACCESS**

#### Edited by:

Jacqueline Batley, The University of Western Australia, Australia

#### Reviewed by:

Tomohiro Ban, Yokohama City University, Japan Umesh K. Reddy, West Virginia State University, United States

\*Correspondence:

Yaxi Liu liuyaxi@sicau.edu.cn <sup>†</sup>These authors have contributed equally to this work

#### Specialty section:

This article was submitted to Plant Breeding, a section of the journal Frontiers in Plant Science

Received: 20 November 2019 Accepted: 29 April 2020 Published: 04 June 2020

#### Citation:

Ma J, Lin Y, Tang S, Duan S, Wang Q, Wu F, Li C, Jiang X, Zhou K and Liu Y (2020) A Genome-Wide Association Study of Coleoptile Length in Different Chinese Wheat Landraces. Front. Plant Sci. 11:677. doi: 10.3389/fpls.2020.00677 From the perspective of wheat yield improvement, the coleoptile is vital for successful crop establishment, and long coleoptile lengths (CLs) are preferred in wheat-growing regions where deep planting is practiced. To determine the genetic basis of CL, we performed a genome-wide association study on a set of 707 Chinese wheat landraces using 18,594 single-nucleotide polymorphisms and 38,678 diversity array technology sequencing markers. We accordingly detected a total of 29 significant markers [ $-\log_{10}^{(P)} > 4.76$ ] distributed on chromosomes 2B, 2D, 3A, 4A, 5A, 6A, 6B, 6D, and 7B. Based on linkage disequilibrium decay distance, we identified a total of 17 quantitative trait loci associated with CL, among which *QCI.sicau-6B.2*, located at 508.17–509.26 Mb on chromosome 6B, was recognized as a novel major locus. We subsequently developed a high-resolution melt marker for *QCI.sicau-6B.2*, which was validated in an  $F_{2:3}$  population. Our findings provide important insights into the genetic mechanisms underlying coleoptile growth and could be applied to marker-assisted wheat selection.

Keywords: candidate gene, coleoptile, genetic variation, Triticum aestivum L., validation

## INTRODUCTION

Successful crop establishment is required for optimal yields and depends upon rapid shoot emergence and leaf area development. The coleoptile plays an important role in the establishment of wheat (*Triticum aestivum* L.), as it transports the young stem and first leaf from the embryo to the soil surface and protects the plumule as it moves through the soil layers. The coleoptile also determines the maximum depth at which wheat seeds can be sown. Approximately 37% of the wheat cropping area in developing countries consists of semi-arid/arid environments (Sinclair, 2011; Sadok, 2017), and compared to regions with higher precipitation levels, wheat is often planted deeper in these arid regions to ensure adequate soil moisture for germination (Schillinger et al., 1998; Mohan et al., 2013). With deep sowing, however, short coleoptiles might limit the ability of the seedling to push the first leaf through the soil surface, resulting in poor establishment. Therefore, wheat cultivars with long coleoptiles are preferable for deep planting.

Quantitative trait loci (QTLs) for coleoptile length (CL) have to date been mapped to wheat chromosomes 1A, 1B, 2B, 2D, 3A, 3B, 3D, 4A, 4B, 4D, 5A, 5B, 5D, 6A, 6B, 7A, and 7B, among

which Rht-B1b (Rht1) and Rht-D1b (Rht2) are two major QTLs on chromosomes 4BS and 4DS, respectively (Rebetzke et al., 2007a, 2014; Spielmeyer et al., 2007; Yu and Bai, 2010; Singh et al., 2015; Li G. et al., 2016). Gibberellic acid (GA)-insensitive Rht1 and Rht2 are associated with reduced cell size and elongation and corresponding reductions in plant height (PH) and CL (Botwright et al., 2001, 2005). Since the advent of the Green Revolution in China, the dwarfing genes Rht1 and Rht2 derived from the cultivar Norin 10 have been used worldwide for dwarf wheat breeding (Worland and Sayers, 1995). Nevertheless, these have hindered wheat breeding for increased CL and reduced PH. In contrast, Rht4, Rht5, Rht7, Rht8, Rht9, Rht12, Rht13, and Rht14 contribute to reducing wheat shoot height but do not affect the CL (Ellis et al., 2004; Botwright et al., 2005), and thus, these genes enable the breeding of dwarf cultivars with long coleoptiles. Moreover, prior to the Green Revolution, wheat landraces had been widely cultivated on a global scale. In China, there are more than 13,000 wheat landraces with relatively high PH (>110 cm) (Dong and Zheng, 2000). Thus, it is reasonable to speculate that the identification of QTLs controlling CL but not influencing PH in Chinese landraces might facilitate the breeding of dwarf cultivars with long coleoptiles.

A genome-wide association study (GWAS) is a powerful approach that can be used to facilitate the identification of QTLs encoding complex traits in plants. Unlike biparental mapping, GWAS exploits the advantages of phenotypic variation and historical recombination in natural populations. It can also overcome the limitations of biparental populations and enable high-resolution mapping (Nordborg and Weigel, 2008; Zhu et al., 2008). To date, GWAS has been widely used to identify QTLs in plants including Arabidopsis (Verslues et al., 2014; Bac-Molenaar et al., 2015), Aegilops tauschii (Liu et al., 2015a,b; Qin et al., 2016; Arora et al., 2017), and maize (Lu et al., 2012; Wen et al., 2014; Yang et al., 2014). In wheat, this technique has also been used to identify QTLs for agronomic traits (Liu Y. et al., 2017; Ogbonnaya et al., 2017), grain yield (Sukumaran et al., 2015; Sehgal et al., 2017), abiotic stress response (Mwadzingeni et al., 2017; Valluru et al., 2017; Lin et al., 2019), disease resistance (Jiang et al., 2015; Maccaferri et al., 2015), and quality-related characteristics (Lin et al., 2017; Liu J. et al., 2018). To date, however, only a single GWAS has been reported for the wheat coleoptile, in which four of eight potentially novel loci were identified for CL (Li G. et al., 2016). Accordingly, performing a GWAS for CL and growth might assist in the identification of novel loci and the development of new wheat varieties with the desired traits. In the present study, we collected 707 wheat landraces from 10 Chinese agro-ecological zones and used these to evaluate CL, in addition to conducting a GWAS on 57,272 markers to identify marker-trait associations and candidate genes.

## MATERIALS AND METHODS

### **Plant Material**

In the present study, we used a core collection of 707 wheat landrace accessions, which were acquired from the following

10 agro-ecological zones in China: the Northern Winter Wheat Zone (NW), the Yellow and Huai River Valleys Facultative Wheat Zone (Y&H), the Middle and Low Yangtze Valleys Autumn-sown Spring Wheat Zone (YTS), the Southern Autumn-sown Spring Wheat Zone (SAS), the Southwestern Autumn-sown Spring Wheat Zone (SWAS), the Northeastern Spring Wheat Zone (NES), the Northern Spring Wheat Zone (NS), the Northwestern Spring Wheat Zone (NWS), the Qinghai-Tibetan Plateau Spring-Winter Wheat Zone (Q&T), and the Xinjiang Winter-Spring Wheat Zone (XJ). Detailed information on these 707 accessions is listed in **Supplementary Table S1**. As a validation population, we used an F<sub>2:3</sub> population derived from the landraces Niqiuzhuan (NQZ) and Huolishao (HLS), which were obtained from an association mapping population and originated in zones III and I, respectively, with the former having a longer coleoptile than the latter.

### Phenotype Evaluation

The evaluation of CL was performed using a blotting-paper germination protocol (Li G. et al., 2016). A completely randomized block experimental design with three replications was used, and for each replication, 30 full, uniformly sized, and undamaged seeds from each accession were evaluated. In brief, two soaked filter papers were placed on wax paper, and the seeds for each accession were then placed in a line on the filter paper. The filter paper and wax paper were subsequently loosely rolled, secured with a rubber band, and arranged vertically in a plastic box. The plastic box was covered with black bags to exclude light and placed in refrigerator at a constant temperature of 4°C for 1 day to break dormancy. Thereafter, the box was transferred to a completely darkened growth chamber incubator at 15°C for 7 days, followed by 7 days at 20°C. Finally, the CLs of the 30 seedlings per accession were measured using a ruler as the distance from the scutellum to the tip of the coleoptile.

## **Statistical Analysis**

Means were determined for the 30 CLs from each independent replicate. To eliminate potential environmental effects, best linear unbiased prediction (BLUP) analyses were conducted among three replicates (Piepho et al., 2008). The means of three replicates (MR1, MR2, and MR3) and the BLUP were used for further analyses. Descriptive statistics and Pearson's correlation coefficients were calculated for the means of each trait in all accessions using IBM SPSS Statistics for Windows v. 20.0 (IBM Corp., Armonk, NY, United States). Data on eight agronomic traits, namely, flowering date (FD), flag leaf length, flag leaf width, peduncle length (PDL), PH, seedling habits (SH), spike length, and tiller number were obtained from a previously published study (Liu Y. et al., 2017). The BLUP values of these eight traits were used to examine the correlations between plant morphology and CL. Analysis of variance was performed using SAS v. 8.1 (SAS Institute Inc., Cary, NC, United States). Broad-sense heritability  $(H^2)$  was evaluated in SAS using the formula H = VG/(VG + VE), where VG is the genetic variance and VE is the environmental variance (Smith et al., 1998).

# Genotyping and Population Structure Analyses

Genomic DNA was extracted from leaf samples of the 707 Chinese wheat landraces and the  $F_{2:3}$  population using the cetyltrimethylammonium bromide method. Genotype by sequence libraries were constructed in 96-plex via the twoenzyme method (Poland et al., 2012) and sequenced using an Illumina HiSeq 2500 system. Single-nucleotide polymorphism (SNP) calling was performed using the TASSEL pipeline (Glaubitz et al., 2014). SNPs were assigned based on the Chinese Spring sequence [Reference sequence (RefSeq) v1.0; the International Wheat Genome Consortium (IWGSC)<sup>1</sup>] to obtain physical distances. SNPs were removed if they could not be assigned to any chromosome. Imputation accuracies were calculated using linkage disequilibrium (LD) K-number neighbor imputation (Money et al., 2015). SNPs with a minor allele frequency (MAF) < 0.05 and missing data > 20% were filtered, yielding a final total of 18,594 SNPs. In addition, we used a set of 38,678 diversity array technology sequencing (DArT-seq) markers that were mapped on the RefSeq v1.0 reference sequence generated in our previous study (Liu Y. et al., 2017).

Population structure was estimated using the STRUCTURE 2.3.4 program implementing model-based Bayesian cluster analysis (Pritchard et al., 2000). Population genetic clusters of K = 1-10 were evaluated using the admixture model with 10,000 replicates for burn-in and 10,000 replicates for Markov Chain Monte Carlo simulation. Five runs per K were performed, with the optimal K-value being determined using STRUCTURE HARVESTER (Earl and vonHoldt, 2012) based on the Evanno method (Evanno et al., 2005). The optimal alignment of the five replicates was determined using CLUMPP (Jakobsson and Rosenberg, 2007), and kinship among the 707 landraces was estimated using TASSEL 5 (Bradbury et al., 2007).

### **Genome-Wide Association Study (GWAS)**

A GWAS was conducted with TASSEL 5 using a compressed mixed linear model with optimal compression and variance component estimation based on previously determined population parameters (Bradbury et al., 2007; Zhang et al., 2010). A model with a Q matrix for the fixed effect and a kinship matrix for the random effect (Yu and Buckler, 2006) was used to identify marker-trait associations. Based on a Bonferroni correction (Yang et al., 2014), we set a threshold of  $-\log_{10} (1/n,$  where n = the number of markers) to detect significant associations between DArT-seq markers and traits. For GWAS, we set a  $-\log_{10}$  threshold of 4.76. Manhattan and quantile-quantile plots were generated using R v. 3.0.3<sup>2</sup>.

### Marker Development and QTL Validation

Based on the IWGSC RefSeq v. 1.0 sequence, extended sequences around significant markers were used to design primers using Primer-BLAST in the National Center for Biotechnology

<sup>1</sup>http://www.wheatgenome.org/

Information<sup>3</sup> database. Sequences of the parental landraces NQZ and HLS were compared, and we searched for SNPs using DNAMAN v. 6.0 (Lynnon BioSoft; Vaudreuil-Dorion, Quebec, Canada). Single-base differences were identified via high-resolution melt marker (HRM) analysis (Wittwer et al., 2003; Wang et al., 2016; Zhou et al., 2016; Liu Y. et al., 2018). The SNPs were converted into HRM markers to track the identified loci by quantitative real-time polymerase chain reaction (qRT-PCR). The HRM markers were designed using Beacon Designer v. 7.9 and evaluated using Oligo v. 6.0 (Zhang and Gao, 2004). The optimal parameters for primer design and the qRT-PCR amplification reactions and the HRM marker polymorphisms, specificities, and sensitivities were determined based on previously described procedures (Wang et al., 2016; Zhou et al., 2016; Liu Y. et al., 2016; Zhou et al., 2016; Liu Y. et al., 2018).

For the development of HRM markers, the CLs of the homozygous NQZ/HLS lines were used to validate significant loci. Genotypes with homozygous alleles from NQZ were designated "AA" and those with homozygous alleles from HLS were identified as "aa." Student's *t*-tests were used to identify the differences in CL between these two classes and the loci effects in the validation population were calculated.

### Identification of Candidate Genes

The significant loci were used to identify candidate genes based on IWGSC RefSeq Annotation v. 1.0. The genes in the two aforementioned genomic regions were annotated using KOBAS v. 3.0 at  $P < 10^{-5}$  with *Arabidopsis thaliana* and *Oryza sativa* as background species (Wu et al., 2006). Identification of the candidate genes for CL was based on functional annotation.

## RESULTS

### **Phenotypic Analysis**

We measured the CLs of 697 landraces and observed significant variation (P < 0.001) among the different genotypes and replicates (**Table 1**). Among these landraces, CL values ranged from 9.26 to 19.40 cm, 9.87 to 20.06 cm, 9.38 to 19.30 cm, and 9.72 to 19.34 cm for MR1, MR2, MR3, and BLUP, respectively (**Figure 1** and **Table 2**). The coefficient of variation ranged from 10.76% to 11.86%, and the  $H^2$  of CL was 95.81%, thereby indicating high heritability (**Table 2**). Among the 10 assessed agro-ecological zones, we found that the CL in the NWS and Q&T zones was higher than that in other zones (**Figure 1**).

# Correlations Between CLs and Agronomic Traits

Correlation analyses performed for associations among CL measurements in the three replicates and BLUP revealed high correlations, with correlation coefficients ranging from 0.889 (between MR1 and MR3) to 0.970 (between MR2 and BLUP) (**Table 3**). To determine whether the CL of landraces is associated with PH and other characteristics, we performed correlation

<sup>&</sup>lt;sup>2</sup>http://www.r-project.org/

<sup>&</sup>lt;sup>3</sup>https://www.ncbi.nlm.nih.gov/tools/primer-blast/index.cgi?LINK\_LOC=BlastHome

Variables	Degree of freedom	Type III sum of squares	Mean squares	F-Value	Significance	
Genotype	696	138958.98	199.65	320.63	***	
Repetition	2	7026.53	3513.26	5642.09	***	
Genotype × Repetition	1392	9971.75	7.16	11.50	***	

TABLE 1 | Analysis of variance for coleoptile length.

\*\*\*Significant at the 0.001 probability level.



analyses between CL and each of the assessed agronomic traits. The correlation coefficients ranged from 0.008 (between MR2 and PH) to 0.235 (between MR3 and FD). The correlation coefficients ranged from 0.012 to 0.227 (**Table 3**). With the exception of PDL and PH, significantly positive correlations (P < 0.05 or P < 0.01) were observed between CL and all other agronomic traits. We found that CL was significantly negatively correlated (P < 0.01) with PDL. No significant correlation was observed between PH and CL. Thus, we could ascertain that certain QTLs or genes, which remain to be identified, are associated with CL but not with PH.

**TABLE 2** | Phenotypic variation and broad-sense heritability  $(H^2)$  of coleoptile length.

Variables	Minimum	Maximum	Mean	$SD^{\dagger}$	CV (%)	H <sup>2</sup> (%)
MR1	9.26	19.40	13.22	1.53	11.57	
MR2	9.87	20.06	13.85	1.54	11.12	
MR3	9.38	19.30	13.07	1.55	11.86	
BLUP	9.72	19.34	13.38	1.44	10.76	95.81

<sup>†</sup>SD, standard deviation; CV, coefficient of variation; H<sup>2</sup>, broad-sense heritability; MR1, means of coleoptile length in repetition 1; MR2, means of coleoptile length in repetition 2; MR3, means of coleoptile length in repetition 3; BLUP, best linear unbiased predictions. Based on BLUP, we also analyzed the association between CL and PH using a quantile method in which the two extreme PH groups were segregated. Genotypes with PH values lower than the 10% quantile were classified as the shorter PH group (SPHG), whereas those with values higher than the 90% quantile were classified in the taller PH group (TPHG) (Valluru et al., 2017), and we accordingly found that the CLs of the SPHG sub-group were not significantly correlated with those of the TPHG sub-group (r = 0.007; P > 0.05), whereas PH was significantly positively correlated with CL in the SPHG sub-group (r = 0.310; P < 0.01).

# Molecular Markers, Population Structure, and Kinship

After filtering markers based on thresholds of MAF < 0.05 and missing data > 20%, we obtained a total of 57,272 polymorphic markers (18,594 SNP and 38,678 DArT-seq markers) for the 707 landraces, which were mapped on a physical reference map (Refseq v1.0). A total of 21,503, 25,365, and 10,404 markers were mapped onto the A, B, and D sub-genomes, respectively (**Table 4**). The marker density for the whole genome was 0.25 Mb/marker and 0.23, 0.20, and 0.38 Mb/marker for the A, B, and D sub-genomes, respectively. The number of markers per chromosome ranged from 476 for Chr. 4D to 4,885 for Chr. 2B, with Chr. 4D and Chr. 2B showing the lowest (1.07 Mb/marker)

TABLE 3 | Pearson correlations between coleoptile length and eight agronomic traits.

MR1 <sup>†</sup>	MR2	MR3	BLUP	FD	FLL	FLW	PDL	SH	SL	TN	PH
1	0.914**	0.889**	0.967**	0.215**	0.145**	0.096*	-0.195**	0.201**	0.227**	0.219**	-0.072
0.914**	1	0.897**	0.970**	0.171**	0.139**	0.131**	-0.186**	0.215**	0.215**	0.156**	-0.063
0.889**	0.897**	1	0.962**	0.128**	0.145**	0.116**	-0.148**	0.209**	0.217**	0.137**	-0.012
0.967**	0.970**	0.962**	1	0.177**	0.148**	0.119**	-0.182**	0.216**	0.227**	0.177**	-0.051
	<b>MR1</b> <sup>†</sup> 1 0.914** 0.889** 0.967**	MR1 <sup>†</sup> MR2   1 0.914**   0.914** 1   0.889** 0.897**   0.967** 0.970**	MR1 <sup>†</sup> MR2 MR3   1 0.914** 0.889**   0.914** 1 0.897**   0.889** 0.897** 1   0.967** 0.970** 0.962**	MR1 <sup>†</sup> MR2 MR3 BLUP   1 0.914** 0.889** 0.967**   0.914** 1 0.897** 0.970**   0.889** 0.897** 1 0.962**   0.967** 0.970** 0.962** 1	MR1 <sup>†</sup> MR2 MR3 BLUP FD   1 0.914** 0.889** 0.967** 0.215**   0.914** 1 0.897** 0.970** 0.171**   0.889** 0.897** 1 0.962** 0.128**   0.967** 0.970** 0.962** 1 0.177**	MR1 <sup>†</sup> MR2 MR3 BLUP FD FLL   1 0.914** 0.889** 0.967** 0.215** 0.145**   0.914** 1 0.897** 0.970** 0.171** 0.139**   0.889** 0.897** 1 0.962** 0.128** 0.145**   0.967** 0.970** 1 0.962** 0.128** 0.145**	MR1 <sup>†</sup> MR2 MR3 BLUP FD FLL FLW   1 0.914** 0.889** 0.967** 0.215** 0.145** 0.096*   0.914** 1 0.897** 0.970** 0.171** 0.139** 0.131**   0.889** 0.897** 1 0.962** 0.128** 0.145** 0.116**   0.967** 0.970** 0.962** 1 0.177** 0.148** 0.119**	MR1 <sup>†</sup> MR2 MR3 BLUP FD FLL FLW PDL   1 0.914** 0.889** 0.967** 0.215** 0.145** 0.096* -0.195**   0.914** 1 0.897** 0.970** 0.171** 0.139** 0.131** -0.186**   0.889** 0.897** 1 0.962** 0.128** 0.145** 0.116*** -0.148**   0.967** 0.970** 0.962** 1 0.177** 0.148** 0.119** -0.182**	MR1 <sup>†</sup> MR2 MR3 BLUP FD FLL FLW PDL SH   1 0.914** 0.889** 0.967** 0.215** 0.145** 0.096* -0.195** 0.201**   0.914** 1 0.897** 0.970** 0.171** 0.139** 0.131** -0.186** 0.215**   0.889** 0.897** 1 0.962** 0.128** 0.145** 0.116** -0.148** 0.209**   0.967** 0.970** 1 0.177** 0.145** 0.110** -0.182** 0.209**   0.967** 0.970** 1 0.177** 0.148** 0.119** -0.182** 0.216**	MR1 <sup>†</sup> MR2 MR3 BLUP FD FLL FLW PDL SH SL   1 0.914** 0.889** 0.967** 0.215** 0.145** 0.096* -0.195** 0.201** 0.227**   0.914** 1 0.897** 0.970** 0.171** 0.139** 0.131** -0.186** 0.215** 0.215**   0.889** 0.897** 1 0.962** 0.128** 0.145** 0.116** -0.148** 0.209** 0.217**   0.967** 0.970** 0.962** 1 0.177** 0.148** 0.119** -0.182** 0.216** 0.227**	MR1 <sup>†</sup> MR2 MR3 BLUP FD FLL FLW PDL SH SL TN   1 0.914** 0.889** 0.967** 0.215** 0.145** 0.096* -0.195** 0.201** 0.227** 0.219**   0.914** 1 0.897** 0.970** 0.171** 0.139** 0.131** -0.186** 0.215** 0.215** 0.156**   0.889** 0.897** 1 0.962** 0.128** 0.145** 0.116** -0.148** 0.209** 0.217** 0.137**   0.967** 0.970** 0.962** 1 0.177** 0.148** 0.119** -0.182** 0.216** 0.227** 0.177**

\*Significant at the 0.05 probability level. \*\*Significant at the 0.01 probability level. <sup>†</sup>MR1, means of coleoptile length in repetition 1; MR2, means of coleoptile length in repetition 2; MR3, means of coleoptile length in repetition 3; BLUP, best linear unbiased predictions; FD, flowering date; FLL, flag leaf length; FLW, flag leaf width; PDL, peduncle length; SH, seedling habit; SL, spike length; TN, tiller number; PH; plant height.

**TABLE 4** | The distribution of molecular marker in the whole genome.

Chr†		Number o	of markers			Map le	Marker density (Mb/marker)					
	A	В	D	Total	А	В	D	Total	Α	В	D	Total
1	2678	3543	1526	7747	593.45	689.28	495.25	1777.98	0.22	0.19	0.32	0.23
2	4048	4885	2678	11611	780.78	801.01	651.69	2233.48	0.19	0.16	0.24	0.19
3	2298	3825	1409	7532	750.76	829.98	615.46	2196.19	0.33	0.22	0.44	0.29
4	2995	1623	476	5094	744.54	673.45	509.85	1927.85	0.25	0.41	1.07	0.38
5	2715	3617	1152	7484	709.76	712.85	566.06	1988.66	0.26	0.20	0.49	0.27
6	2969	3878	1552	8399	618.00	720.95	473.52	1812.47	0.21	0.19	0.31	0.22
7	3800	3994	1611	9405	736.69	750.59	638.66	2125.94	0.19	0.19	0.40	0.23
All	21503	25365	10404	57272	4933.96	5178.11	3950.49	14062.56	0.23	0.20	0.38	0.25

<sup>†</sup>Chr, chromosome.

and highest (0.16 Mb/marker) marker densities, respectively. Based on the delta *K* model, the 707 landraces were divided into five sub-groups comprising 207, 185, 128, 105, and 82 landraces, respectively, as follows: sub-group 1 included most of the landraces from NW and Y&H and all of the landraces from XJ; sub-group 2 included most of the landraces from YTS and SAS; sub-group 3 included most of the landraces from SWAS and some of the landraces from Y&H and YTS; subgroup 4 included most of the landraces from IX and Q&T and some of the landraces from SWAS, NS, and NWS; subgroup 5 was a more diverse group that included most of the landraces from NEW and some of the landraces from Y&H, YTS, SWAS, and Q&T.

### Significant Markers Associated With CL

A GWAS was performed on MR1, MR2, MR3, and BLUP, using a total of 57,272 DArT-seq markers, with a  $-\log_{10} {}^{(P)}$  value > 4.76  $[-\log_{10} {}^{(P)} = -\log_{10} {}^{(1/57,272)}]$  set as the threshold (Yang et al., 2014), to identify significant marker-trait associations. We accordingly detected a total of 29 significant markers for CL, with the amount of phenotypic variation explained (PVE) ranging from 2.70 to 10.64% (**Figure 2** and **Table 5**). These markers were found to be distributed on chromosomes 2A, 2B, 2D, 3A, 3B, 4D, 5A, 6B, 7A, and 7D. Based on the LD decay distance, significant markers falling within a 5.98-Mb region were considered to represent a single QTL (Lin et al., 2019). In total, we detected 17 QTLs for CL, among which *QCLsicau-6B.2*, located at 508.17–509.26 Mb on Chr. 6B, was considered a major locus and was detected in all three replicates and BLUP, with PVE values ranging from 3.10 to 8.17%.

# Development of an HRM Marker to Validate *QCI.sicau-6B.2*

A newly designed primer, HRM, was used to track QTL-6B-2, the information on which is presented in **Supplementary Table S2**. This marker was polymorphic in the  $F_{2:3}$  population, in which we detected 103 individuals with the homozygous "AA" genotype and 106 with the homozygous "aa" genotype. We found that the CL of the homozygous "AA" lines ranged from 11.53 to 16.11 cm, whereas that of the homozygous "aa" lines ranged from 9.11 to 14.81 cm (**Figure 3**). Furthermore, the average CL of NQZ landrace lines with homozygous genes was significantly greater than that of HLS lines with homozygous genes, and the effect was 9.48%.

# Predicted Genes Located in the Genomic Region of *QCI.sicau-6B.2*

The major QTL *QCl.sicau-6B.2* encompassed four significant markers, which were distributed over a 1.09-Mb region on chromosome 6B (Chr. 6B: 508.17–509.26 Mb) that contained 15 predicted genes (**Supplementary Table S3**). A gene ontology analysis indicated three genes with no specific annotation. Twelve of these genes were found to be homologous to genes occurring in *Arabidopsis*, including the four previously reported genes *G3Pp4* (Kawai et al., 2014), *WRKY12* (Wang et al., 2010; Sanchez et al., 2012; Li W. et al., 2016), *CYCP4;1* (Acosta et al., 2004), and *petN* (Sato et al., 1999), whereas 13 of the predicted genes were found to have rice homologs, including the three reported genes *P0491E01.9* (Kikuchi et al., 2003), *OrsajM\_p08* (Hiratsuka et al., 1989), and ycf6 (Hiratsuka et al., 1989; Boeckmann et al., 2003).



# DISCUSSION

We detected large phenotypic variations in the CL of 697 Chinese wheat landraces, with values ranging from 9.26 to 20.06 cm and a mean value of  $\sim$ 13 cm. Consistently, based on an analysis of a worldwide collection of 893 wheat accessions. Li G. et al. (2016) obtained a mean CL of 13.33 cm. Furthermore, we found that landraces tend to have longer CLs than those previously determined for breeding lines or cultivars (Li G. et al., 2016), whereas in comparison, the CLs of an RIL population derived from a cross between the Chinese landrace Wangshuibai and the United States cultivar Wheaton ranged from 7.7 to 15.9 cm (Yu and Bai, 2010). Moreover, consistent with the findings of the present study, high correlations among experiments/repetitions and high heritabilities were detected in this RIL population and the aforementioned 893 worldwide wheat accessions (Yu and Bai, 2010; Li G. et al., 2016). Such high phenotypic variation and heritability was conducive to obtaining a high resolution for CL using GWAS.

The LD decay distance for Chinese wheat landraces was 5.98 Mb for the whole genome and 5.83, 5.28, and 6.00 Mb for A, B, and D sub-genomes, respectively (Lin et al., 2019), indicating that a large number of markers is required to achieve high resolution. We obtained a combined total of 57,272 SNP and DArT-seq markers with a marker density of 0.25 Mb/marker for the whole panel and sufficient SNP density for high-resolution mapping. Comparatively, Li G. et al. (2016) obtained an LD decay distance of 10 cM for the whole genome among the set of 893 global wheat accessions they examined and used 4,716 SNPs for further GWAS of CL. Previous studies on wheat have obtained LD decay values ranging from 1.5 to 15 cM, and found that the LD of the D sub-genome was

higher than that of the A and B sub-genomes (Chao et al., 2010; Hao et al., 2011; Sukumaran et al., 2015; Li G. et al., 2016; Rahimi et al., 2019), which is consistent with the findings of the present study. We also found that the diversity panel could be divided into three sub-groups, which were largely consistent with geographical origins. Similarly, previous studies on wheat have reported that the characterization of sub-groups divided by Bayesian cluster analysis was largely consistent with geographical origins and pedigrees (Sukumaran et al., 2015; Li G. et al., 2016; Liu J. et al., 2017). In the present study, we considered population structure as a fixed-effect to reduce false positive errors and adopted MLM adjusted by population structure and kinship implemented in TASSEL v5.0 for association analysis.

Based on GWAS performed on 57,272 polymorphic markers associated with wheat CL, we detected a total of 29 markers that were significantly associated with this trait (Figure 2 and Table 5). The QTLs we thus identified for CL are comparable with those previously identified. Among these, the locus QCl.sicau-6B.2, which was associated with four significant markers (B18505 and B18506, SNP14693, and A17510) in a 1.09-Mb genomic region (508.17-509.26 Mb) on Chr. 6B, was found to have the strongest effect on CL. Previous studies have reported 87 loci associated with CL (Supplementary Table S4), three of which are located on chromosome 6B (Rebetzke et al., 2007a, 2014; Singh et al., 2015). The markers closest to these QTLs were barc178, gwm219, and P35/M39-9, the first two of which are located at 664.63 and 674.84 Mb, respectively, and thus relatively remote from the loci identified in the present study. However, the P35/M39-9 probe sequence was not published and cannot be compared based on the Chinese Spring physical map. According to Rebetzke et al. (2007a), the QTL interval of P35/M39-9 was close to that of

QTL Name	Marker	$\mathbf{Chr}^{\dagger}$	Position (Mb)		-LO	G <sub>10</sub> <sup>(P)</sup>			PV	E (%)	
				MR1	MR2	MR3	BLUP	MR1	MR2	MR3	BLUP
QCI.sicau-2A	B32929	2A	764.66	_	6.03	5.18	6.07	_	4.10	3.57	4.17
	SNP3882	2A	764.66	6.43	6.32	6.39	7.36	3.75	3.68	3.74	4.36
QCI.sicau-2B.1	A30893	2B	39.41	4.93	-	-	-	2.80	-	-	-
QCI.sicau-2B.2	A30968	2B	47.85	-	-	4.96	-	-	-	2.97	-
QCI.sicau-2D	SNP6484	2D	651.77	-	5.23	-	5.01	-	2.96	-	2.82
QCI.sicau-3A.1	A48978	ЗA	404.69	-	5.20	-	-	-	3.13	-	-
	SNP6783	ЗA	404.69	-	4.87	-	-	-	2.73	-	-
QCI.sicau-3A.2	SNP6979	ЗA	700.22	5.28	-	-	5.13	3.00	-	-	2.90
	B11794	ЗA	700.22	5.34	-	-	5.16	3.03	-	-	2.92
	B11790	ЗA	700.22	5.29	-	-	5.13	3.00	-	-	2.90
	B11795	ЗA	700.22	5.26	-	_	5.11	2.99	-	_	2.90
QCI.sicau-3B.1	B10892	3B	755.81	5.23	5.20	-	5.39	3.08	3.07	-	3.19
	B10891	3B	755.81	4.85	4.82	-	5.00	2.83	2.82	-	2.94
	SNP8235	3B	755.81	5.44	5.34	-	5.58	3.10	3.04	-	3.19
QCI.sicau-3B.2	SNP8294	3B	770.47	5.32	5.67	-	5.76	3.02	3.25	-	3.31
QCI.sicau-3B.3	A11280	3B	813.01	-	-	4.80	-	-	-	2.84	-
QCI.sicau-4D.1	A13215	4D	319.09	-	-	4.92	-	-	-	2.99	-
QCI.sicau-4D.2	A39373	4D	498.96	-	5.24	-	-	-	3.02	-	-
QCI.sicau-5A	SNP11051	5A	564.56	-	4.92	-	4.89	-	2.76	-	2.75
	B14507	5A	564.56	5.02	5.17	-	5.13	2.86	2.94	-	2.92
	SNP11058	5A	566.55	-	4.92	-	4.89	-	2.76	-	2.75
QCI.sicau-6B.1	A48008	6B	168.77	5.85	-	5.00	5.50	3.52	-	3.04	3.32
QCI.sicau-6B.2	B18505	6B	508.17	10.53	9.30	12.02	11.49	6.93	6.07	8.17	7.67
	A17510	6B	508.17	6.44	5.36	6.19	6.42	3.85	3.10	3.71	3.82
	SNP14693	6B	508.17	7.54	6.54	7.74	7.75	4.48	3.82	4.63	4.62
	B18506	6B	509.26	7.97	7.94	10.19	9.33	5.08	5.09	6.80	6.09
QCI.sicau-6B.3	SNP14819	6B	628.06	-	-	5.07	-	-	-	2.88	-
QCI.sicau-7A	A19567	7A	536.82	-	-	4.77	-	-	-	2.76	-
QCI.sicau-7D	A888	7D	23.52	-	-	-	4.79	-	-	-	2.90

#### TABLE 5 | List of the significant markers for coleoptile length.

<sup>†</sup>Chr, chromosome; PVE, phenotypic variation explained; MR2, means of coleoptile length in repetition 2; MR3, means of coleoptile length in repetition 3; BLUP, best linear unbiased predictions.

barc178 and gwm219 on their map. Using a MAGIC population, the two significant SNPs wsnp\_Ex\_rep\_c71537\_70252046 and wsnp\_Ex\_rep\_c70767\_69655253 were detected on chromosome 6B at 690.73 and 692.78 Mb, respectively (Rebetzke et al., 2014). Again, however, these are located at some distance from the loci detected in the present study. The QTL qCL.6B.1 (with flanking markers gpw1017 and barc79) has previously been mapped to chromosome 6B for CL in an RIL population (Singh et al., 2015). The marker gpw1017 was not uniquely located on the Chinese Spring physical map, and thus, we compared the marker barc79 located at 582.38 Mb with the locus identified in the present study. The distance between barc79 and our locus was >70 Mb, thereby suggesting a novel locus for CL on chromosome 6B. We subsequently designed an HRM marker for this locus, which was validated in an  $F_{2:3}$  population, with an effect of 9.48% for CL. The gene TraesCS6B01G282600, located at 508.17-509.26 Mb on Chr. 6B, is a homolog of CYCP4;1 (cyclin p4;1), a gene associated with cell division and differentiation that is strongly expressed in Arabidopsis hypocotyls (Acosta et al., 2004). Accordingly, we speculate that

*TraesCS6B01G282600* might be a candidate gene associated with coleoptile traits.

In the present study, we found that with the exception of PH, there were significant correlations between CL and all other assessed agronomic traits (Table 3). However, we detected a significantly positive correlation between PH and CL in the SPHG sub-group (r = 0.310, P < 0.01), but not in the TPHG sub-group (P > 0.05), thereby indicating that CL is independent of PH among the TPHG landraces, which is consistent with the findings of previous studies. For example, in a comparison of nine semidwarf and seven tall wheat genotypes sown at depths, Rebetzke et al. (2007b) detected no significant correlation between CL and PH among the tall wheat varieties (Rebetzke et al., 2007b). In contrast, a positive correlation has been observed between PH and CL in wheat bearing the GA-insensitive dwarfing genes Rht1 and Rht2 (Rebetzke et al., 2007b; Liatukas and Ruzgas, 2011), which have the effect of suppressing CL and PH (Ellis et al., 2004). In the present study, we identified a total of 229 significant markers for the eight assessed agronomic traits based on GWAS (Supplementary Table S5). Interestingly, a genomic



\*\*\*

Allele

FIGURE 3 | Box plots of coleoptile length (CL) in a *F*<sub>2:3</sub> population. "AA" and "aa" represents homozygous alleles from D519 and D55, respectively.

0

aa

### CONCLUSION

The coleoptile of wheat is essential for the successful establishment of this crop. To identify the genetic basis of CL, we performed a GWAS based on 57,272 SNP and DArT-seq markers for this trait, and accordingly identified a total of 29 significant markers. Among these, a novel locus was identified

on chromosome 6B and validated in an  $F_{2:3}$  population. Based on annotation, we suggest that *TraesCS6B01G282600*, which is associated with cell division and differentiation, is a candidate gene for this locus. In our future research, we intend to use the validated locus in molecular-assisted breeding to develop wheat cultivars with long coleoptiles. The candidate gene will also be validated by qRT-PCR and applied to transgenic tests.

## DATA AVAILABILITY STATEMENT

The datasets GENERATED for this study can be found in the EVA (https://www.ebi.ac.uk/eva/); Project: PRJEB38030 and Analyses: ERZ1309082.

### **AUTHOR CONTRIBUTIONS**

JM and YuL performed the phenotypic evaluation, conducted data analysis, and drafted the manuscript. ST, SD, QW, FW, and CL performed the phenotypic evaluation and helped with data analysis. XJ helped to draft the manuscript. KZ performed part of the statistical analyses. YaL designed and coordinated the study and revised the manuscript. All authors read and approved the final draft of the manuscript.

### FUNDING

This study was supported by the National Key Research and Development Program of China (2016YFD0101004 and 2017YFD0100900), the National Natural Science Foundation of China (31771794), the Fundamental Research Funds for the Central Universities (2019TC214), the outstanding Youth Foundation of the Department of Science and Technology of Sichuan Province (2016JQ0040), the Key Technology Research and Development Program of the Department of Science and Technology of Sichuan Province (2016NZ0057), and the International Science and Technology Cooperation Program of the Bureau of Science and Technology of Chengdu China (No. 2015DFA306002015-GH0300008-HZ).

# ACKNOWLEDGMENTS

The authors thank Dr. Lihui Li (Chinese Academy of Agricultural Sciences, China) for providing plant materials.

### SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fpls.2020.00677/ full#supplementary-material

8

9

(cm) 14

10

AA

\*\*\*Significant at the 0.001 probability level.

2 CL

### REFERENCES

- Acosta, J. T., de Almeida Engler, J., Raes, J., Magyar, Z., De Groodt, R., Inzé, D., et al. (2004). Molecular characterization of *Arabidopsis* PHO80-like proteins, a novel class of CDKA;1-interacting cyclins. *Cell. Mol. Life Sci.* 61, 1485–1497.
- Arora, S., Singh, N., Kaur, S., Bains, N. S., Uauy, C., Poland, J., et al. (2017). Genome-wide association study of grain architecture in wild wheat *Aegilops tauschii. Front. Plant Sci.* 8:886. doi: 10.3389/fpls.2017.00886
- Bac-Molenaar, J. A., Vreugdenhil, D., Granier, C., and Keurentjes, J. J. (2015). Genome-wide association mapping of growth dynamics detects time-specific and general quantitative trait loci. J. Exp. Bot. 66, 5567–5580. doi: 10.1093/jxb/ erv176
- Boeckmann, B., Bairoch, A., Apweiler, R., Blatter, M. C., Estreicher, A., Gasteiger, E., et al. (2003). The SWISS-PROT protein knowledgebase and its supplement TrEMBL in 2003. *Nucleic Acids Res.* 31, 365–370. doi: 10.1093/nar/ gkg095
- Botwright, T., Rebetzke, G., Condon, T., and Richards, R. (2001). The effect of rht genotype and temperature on coleoptile growth and dry matter partitioning in young wheat seedlings. *Funct. Plant Biol.* 28, 417–423. doi: 10.1071/PP01010
- Botwright, T. L., Rebetzke, G. J., Condon, A. G., and Richards, R. A. (2005). Influence of the gibberellin-sensitive Rht8 dwarfing gene on leaf epidermal cell dimensions and early vigour in wheat (*Triticum aestivum L.*). Ann. Bot. 95, 631–639. doi: 10.1093/aob/mci069
- Bradbury, P. J., Zhang, Z., Kroon, D. E., Casstevens, T. M., Ramdoss, Y., and Buckler, E. S. (2007). TASSEL: software for association mapping of complex traits in diverse samples. *Bioinformatics* 23, 2633–2635. doi: 10.1093/ bioinformatics/btm308
- Chao, S., Dubcovsky, J., Dvorak, J., Luo, M. C., Baenziger, S. P., Matnyazov, R., et al. (2010). Population-and genome-specific patterns of linkage disequilibrium and SNP variation in spring and winter wheat (*Triticum aestivum* L.). *BMC Genomics* 11:727. doi: 10.1186/1471-2164-11-727
- Dong, Y. C., and Zheng, D. S. (2000). *Genetic Resources Of Wheat In China*. Beijing: China Agriculture Press.
- Earl, D. A., and vonHoldt, B. M. (2012). Structure HARVESTER: a website and program for visualizing structure output and implementing the Evanno method. *Conserv. Genet. Resour.* 4, 359–361. doi: 10.1007/s12686-011-9548-7
- Ellis, M. H., Rebetzke, G. J., Chandler, P., Bonnett, D., Spielmeyer, W., and Richards, R. A. (2004). The effect of different height reducing genes on the early growth of wheat. *Funct. Plant Biol.* 31, 583–589. doi: 10.1071/FP03207
- Evanno, G., Regnaut, S., and Goudet, J. (2005). Detecting the number of clusters of individuals using the software structure: a simulation study. *Mol. Ecol.* 14, 2611–2620. doi: 10.1111/j.1365-294X.2005.02553.x
- Glaubitz, J. C., Casstevens, T. M., Lu, F., Harriman, J., Elshire, R. J., Sun, Q., et al. (2014). TASSEL-GBS: a high capacity genotyping by sequencing analysis pipeline. *PLoS One* 9:e90346. doi: 10.1371/journal.pone.0090346
- Hao, C., Wang, L., Ge, H., Dong, Y., and Zhang, X. (2011). Genetic diversity and linkage disequilibrium in Chinese bread wheat (*Triticum aestivum* L.) revealed by SSR markers. *PLoS One* 6:e17279. doi: 10.1371/journal.pone.0017279
- Hiratsuka, J., Shimada, H., Whittier, R., Ishibashi, T., Sakamoto, M., Mori, M., et al. (1989). The complete sequence of the rice (*Oryza sativa*) chloroplast genome: intermolecular recombination between distinct tRNA genes accounts for a major plastid DNA inversion during the evolution of the cereals. *Mol. Gen. Genet.* 217, 185–194. doi: 10.1007/BF02464880
- Jakobsson, M., and Rosenberg, N. A. (2007). CLUMPP: a cluster matching and permutation program for dealing with label switching and multimodality in analysis of population structure. *Bioinformatics* 23, 1801–1806. doi: 10.1093/ bioinformatics/btm233
- Jiang, Y., Zhao, Y., Rodemann, B., Plieske, J., Kollers, S., Korzun, V., et al. (2015). Potential and limits to unravel the genetic architecture and predict the variation of Fusarium head blight resistance in European winter wheat (*Triticum aestivum L.*). *Heredity* 114, 318–326. doi: 10.1038/hdy.2014.104
- Kawai, H., Ishikawa, T., Mitsui, T., Kore-eda, S., Yamada-Kawai, M., and Ohnishi, J. I. (2014). Arabidopsis glycerol-3-phosphate permease 4 is localized in the plastids and involved in the accumulation of seed oil. Plant Biotechnol. 30, 159–165. doi: 10.5511/plantbiotechnology.14.0222a
- Kikuchi, S., Satoh, K., Nagata, T., Kawagashira, N., Doi, K., Kishimoto, N., et al. (2003). Collection, mapping, and annotation of over 28,000 cDNA clones from japonica rice. *Science* 301, 376–379. doi: 10.1126/science.1081288

- Li, G., Bai, G., Carver, B. F., Elliott, N. C., Bennett, R. S., Wu, Y., et al. (2016). Genome-wide association study reveals genetic architecture of coleoptile length in wheat. *Theor. Appl. Genet.* 130, 1–11. doi: 10.1007/s00122-016-2820-1
- Li, W., Wang, H., and Yu, D. (2016). The Arabidopsis WRKY transcription factors WRKY12 and WRKY13 oppositely regulate flowering under short-day conditions. Mol. Plant 9, 1492–1503. doi: 10.1016/j.molp.2016.08.003
- Liatukas, Z., and Ruzgas, V. (2011). Coleoptile length and plant height of modern tall and semi-dwarf European winter wheat varieties. *Acta Soc. Bot. Pol.* 80, 197–203. doi: 10.5586/asbp.2011.018
- Lin, Y., Liu, S., Liu, Y., Liu, Y., Chen, G., Xu, J., et al. (2017). Genome-wide association study of pre-harvest sprouting resistance in Chinese wheat founder parents. *Genet. Mol. Biol.* 40, 620–629. doi: 10.1590/1678-4685-gmb-2016-0207
- Lin, Y., Yi, X., Tang, S., Chen, W., Yang, X., Jiang, X., et al. (2019). Dissection of phenotypic and genetic variation of drought-related traits in diverse Chinese wheat landraces. *Plant Genome* 12, 1–14. doi: 10.3835/plantgenome2019.03. 0025
- Liu, J., Feng, B., Xu, Z., Fan, X., Jiang, F., Jin, X., et al. (2018). A genome-wide association study of wheat yield and quality-related traits in southwest China. *Mol. Breed.* 38:1. doi: 10.1007/s11032-017-0759-9
- Liu, Y., Tao, Y., Wang, Z., Guo, Q., Wu, F., Yang, X., et al. (2018). Identification of QTL for flag leaf length in common wheat and their pleiotropic effects. *Mol. Breed.* 38:11. doi: 10.1007/s11032-017-0766-x
- Liu, J., He, Z., Rasheed, A., Wen, W., and Yan, J. (2017). Genome-wide association mapping of black point reaction in common wheat (*Triticum aestivum* L.). BMC Plant Biol. 17:220. doi: 10.1186/s12870-017-1167-3
- Liu, Y., Lin, Y., Gao, S., Li, Z., Ma, J., Deng, M., et al. (2017). A genome-wide association study of 23 agronomic traits in Chinese wheat landraces. *Plant J.* 91, 861–873. doi: 10.1111/tpj.13614
- Liu, Y., Wang, L., Deng, M., Li, Z., Lu, Y., Wang, J., et al. (2015a). Genome-wide association study of phosphorus-defciency-tolerance traits in *Aegilops tauschii*. *Theor. Appl. Genet.* 128, 2203–2212. doi: 10.1007/s00122-015-2578-x
- Liu, Y., Wang, L., Mao, S., Liu, K., Lu, Y., Wang, J., et al. (2015b). Genomewide association study of 29 morphological traits in *Aegilops tauschii. Sci. Rep.* 5:15562. doi: 10.1038/srep15562
- Lu, Y., Xu, J., Yuan, Z., Hao, Z., Xie, C., Li, X., et al. (2012). Comparative LD mapping using single SNPs and haplotypes identifies QTL for plant height and biomass as secondary traits of drought tolerance in maize. *Mol. Breed.* 30, 407–418. doi: 10.1007/s11032-011-9631-5
- Maccaferri, M., Zhang, J., Bulli, P., Abate, Z., Chao, S., Cantu, D., et al. (2015). A genome-wide assocition study of resistance to stripe rust (*Puccinia striiformisf.* sp. tritici) in a worldwide collection of hexaploid spring wheat (*Triticum aestivum* L.). G3 5, 449–465. doi: 10.1534/g3.114.014563
- Mohan, A., Schillinger, W. F., and Gill, K. S. (2013). Wheat seedling emergence from deep planting depths and its relationship with coleoptile length. *PLoS One* 8:e73314. doi: 10.1371/journal.pone.0073314
- Money, D., Gardner, K., Migicovsky, Z., Schwaninger, H., Zhong, G. Y., and Myles, S. (2015). LinkImpute: fast and accurate genotype imputation for non-model organisms. G3 5:2383. doi: 10.1534/g3.115.021667
- Mwadzingeni, L., Shimelis, H., Rees, D. J. G., and Tsilo, T. J. (2017). Genomewide association analysis of agronomic traits in wheat under drought-stressed and non-stressed conditions. *PLoS One* 12:e0171692. doi: 10.1371/journal.pone. 0171692
- Nordborg, M., and Weigel, D. (2008). Next-generation genetics in plants. *Nature* 456:720. doi: 10.1038/nature07629
- Ogbonnaya, F. C., Rasheed, A., Okechukwu, E. C., Jighly, A., Makdis, F., Wuletaw, T., et al. (2017). Genome-wide association study for agronomic and physiological traits in spring wheat evaluated in a range of heat prone environments. *Theor. Appl. Genet.* 130, 1–17. doi: 10.1007/s00122-017-2927-z
- Piepho, H. P., Möhring, J., Melchinger, A. E., and Büchse, A. (2008). BLUP for phenotypic selection in plant breeding and variety testing. *Euphytica* 161, 209–228. doi: 10.1007/s10681-007-9449-8
- Poland, J. A., Brown, P. J., Sorrells, M. E., and Jannink, J. L. (2012). Development of high-density genetic maps for barley and wheat using a novel twoenzyme genotyping-by-sequencing approach. *PLoS One* 7:e32253. doi: 10.1371/journal. pone.0032253
- Pritchard, J. K., Stephens, M., and Donnelly, P. (2000). Inference of population structure using multilocus genotype data. *Genetics* 155, 945–959.

- Qin, P., Lin, Y., Hu, Y., Liu, K., Mao, S., Li, Z., et al. (2016). Genome-wide association study of drought-related resistance traits in *Aegilops tauschii. Genet. Mol. Biol.* 39, 398–407. doi: 10.1590/1678-4685-GMB-2015-0232
- Rahimi, Y., Bihamta, M. R., Taleei, A., Alipour, H., and Ingvarsson, P. K. (2019). Genome-wide association study of agronomic traits in bread wheat reveals novel putative alleles for future breeding programs. *BMC Plant Biol.* 19:541. doi: 10.1186/s12870-019-2165-4
- Rebetzke, G. J., Ellis, M. H., Bonnett, D. G., and Richards, R. A. (2007a). Molecular mapping of genes for coleoptile growth in bread wheat (*Triticum aestivum* L.). *Theor. Appl. Genet.* 114, 1173–1183. doi: 10.1007/s00122-007-0509-1
- Rebetzke, G. J., Richards, R. A., Fettell, N. A., Long, M., Condon, A. G., Forrester, R. I., et al. (2007b). Genotypic increases in coleoptile length improves stand establishment, vigour and grain yield of deep-sown wheat. *Field Crops Res.* 100, 10–23. doi: 10.1016/j.fcr.2006.05.001
- Rebetzke, G. J., Verbyla, A. P., Verbyla, K. L., Morell, M. K., and Cavanagh, C. R. (2014). Use of a large multiparent wheat mapping population in genomic dissection of coleoptile and seedling growth. *Plant Biotechnol. J.* 12, 219–230. doi: 10.1111/pbi.12130
- Sadok, W. (2017). "Wheat," in Water-Conservation Traits To Increase Crop Yields In Water-Deficit Environments: Case Studies, ed. T. R. Sinclair (Cham: Springer International Publishing), 85–92.
- Sanchez, P., Nehlin, L., and Greb, T. (2012). From thin to thick: major transitions during stem development. *Trends in Plant Sci.* 17, 113–121. doi: 10.1016/j. tplants.2011.11.004
- Sato, S., Nakamura, Y., Kaneko, T., Asamizu, E., and Tabata, S. (1999). Complete structure of the chloroplast genome of *Arabidopsis thaliana*. DNA Res. 6, 283–290. doi: 10.1093/dnares/6.5.283
- Schillinger, W. F., Donaldson, E., Allan, R. E., and Jones, S. S. (1998). Winter wheat seedling emergence from deep sowing depths. *Agron. J.* 90, 582–586. doi: 10.2134/agronj1998.00021962009000050002x
- Sehgal, D., Autrique, E., Singh, R., Ellis, M., Singh, S., and Dreisigacker, S. (2017). Identification of genomic regions for grain yield and yield stability and their epistatic interactions. *Sci. Rep.* 7:41578. doi: 10.1038/srep41578
- Sinclair, T. R. (2011). Challenges in breeding for yield increase for drought. *Trends Plant Sci.* 16, 289–293. doi: 10.1016/j.tplants.2011.02.008
- Singh, K., Shukla, S., Kadam, S., Semwal, V. K., Singh, N. K., and Khanna-Chopra, R. (2015). Genomic regions and underlying candidate genes associated with coleoptile length under deep sowing conditions in a wheat RIL population. *J. Plant Biochem. Biotechnol.* 24, 324–330. doi: 10.1007/s13562-014-0277-3
- Smith, S., Kuehl, R., Ray, I., Hui, R., and Soleri, D. (1998). Evaluation of simple methods for estimating broad-sense heritability in stands of randomly planted genotypes. *Crop Sci.* 38, 1125–1129. doi: 10.2135/cropsci1998. 0011183X003800050003x
- Spielmeyer, W., Hyles, J., Joaquim, P., Azanza, F., Bonnett, D., Ellis, M. E., et al. (2007). A QTL on chromosome 6A in bread wheat (*Triticum aestivum*) is associated with longer coleoptiles, greater seedling vigour and final plant height. *Theor. Appl. Genet.* 115, 59–66. doi: 10.1007/s00122-007-0540-2
- Sukumaran, S., Dreisigacker, S., Lopes, M., Chavez, P., and Reynolds, M. P. (2015). Genome-wide association study for grain yield and related traits in an elite spring wheat population grown in temperate irrigated environments. *Theor. Appl. Genet.* 128, 353–363. doi: 10.1007/s00122-014-2435-3
- Valluru, R., Reynolds, M. P., Davies, W. J., and Sukumaran, S. (2017). Phenotypic and genome-wide association analysis of spike ethylene in diverse wheat genotypes under heat stress. *New Phytol.* 214, 271–283. doi: 10.1111/nph.14367
- Verslues, P. E., Lasky, J. R., Juenger, T. E., and Liu, T. W. (2014). Genome-wide association mapping combined with reverse genetics identifies new effectors of

low water potential-induced proline accumulation in *Arabidopsis*. *Plant Physiol*. 164, 144–159. doi: 10.1104/pp.113.224014

- Wang, H., Avci, U., Nakashima, J., Hahn, M. G., Chen, F., and Dixon, R. A. (2010). Mutation of WRKY transcription factors initiates pith secondary wall formation and increases stem biomass in dicotyledonous plants. *Proc. Natl. Acad. Sci. U.S.A.* 107, 22338–22343. doi: 10.1073/pnas.101643 6107
- Wang, Z., Liu, Y., Shi, H., Mo, H., Wu, F., Lin, Y., et al. (2016). Identification and validation of novel low-tiller number QTL in common wheat. *Theor. Appl. Genet.* 129, 603–612. doi: 10.1007/s00122-015-2652-4
- Wen, W., Li, D., Li, X., Gao, Y., Li, W., Li, H., et al. (2014). Metabolome-based genome-wide association study of maize kernel leads to novel biochemical insights. *Nat. Commun.* 5:3438. doi: 10.1038/ncomms4438
- Wittwer, C. T., Reed, G. H., Gundry, C. N., Vandersteen, J. G., and Pryor, R. J. (2003). High-resolution genotyping by amplicon melting analysis using LCGreen. *Clin. Chem.* 49, 853–860. doi: 10.1373/49.6.853
- Worland, A. J., and Sayers, E. J. (1995). *Rht1 (B. dw)* an alternative allelic variant for breeding semi-dwarf wheat varieties. *Plant Breed*. 114, 397–400. doi: 10.1111/j. 1439-0523.1995.tb00819.x
- Wu, J., Mao, X., Cai, T., Luo, J., and Wei, L. (2006). KOBAS server: a web-based platform for automated annotation and pathway identification. *Nucleic Acids Res.* 34, W720–W724. doi: 10.1093/nar/gkl167
- Yang, N., Lu, Y., Yang, X., Huang, J., Zhou, Y., Ali, F., et al. (2014). Genome wide association studies using a new nonparametric model reveal the genetic architecture of 17 agronomic traits in an enlarged maize association panel. *PLoS Genet.* 10:e1004573. doi: 10.1371/journal.pgen.1004573
- Yu, J., and Buckler, E. S. (2006). Genetic association mapping and genome organization of maize. *Curr. Opin. Biotechnol.* 17, 155–160. doi: 10.1016/j. copbio.2006.02.003
- Yu, J. B., and Bai, G. H. (2010). Mapping quantitative trait loci for long coleoptile in Chinese wheat landrace Wangshuibai. *Crop Sci.* 50, 43–50. doi: 10.2135/ cropsci2009.02.0065
- Zhang, X. Y., and Gao, Y. N. (2004). To design PCR primers with Oligo 6 and Primer Premier 5. *Bioinformatiocs* 4:3.
- Zhang, Z., Ersoz, E., Lai, C. Q., Todhunter, R. J., Tiwari, H. K., Gore, M. A., et al. (2010). Mixed linear model approach adapted for genome-wide association studies. *Nat. Genet.* 42, 355–360. doi: 10.1038/ng.546
- Zhou, H., Liu, S., Liu, Y., Liu, Y., You, J., Deng, M., et al. (2016). Mapping and validation of major quantitative trait loci for kernel length in wild barley (*Hordeum vulgare ssp. spontaneum*). BMC Genet. 17:130. doi: 10.1186/1471-2164-11-130
- Zhu, C., Gore, M., Buckler, E. S., and Yu, J. (2008). Status and prospects of association mapping in plants. *Plant Genome* 1, 5–20. doi: 10.3835/ plantgenome2008.02.0089

**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.

Copyright © 2020 Ma, Lin, Tang, Duan, Wang, Wu, Li, Jiang, Zhou and Liu. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.