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MULTIPARENTAL POPULATIONS

Genetic Architecture of Chilling Tolerance in Sorghum Dissected with a Nested Association Mapping Population

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ABSTRACT Dissecting the genetic architecture of stress tolerance in crops is critical to understand and improve adaptation. In temperate climates, early planting of chilling-tolerant varieties could provide longer growing seasons and drought escape, but chilling tolerance (<15°) is generally lacking in tropical-origin crops. Here we developed a nested association mapping (NAM) population to dissect the genetic architecture of early-season chilling tolerance in the tropical-origin cereal sorghum *(Sorghum bicolor* [L.] Moench). The NAM resource, developed from reference line BTx623 and three chilling-tolerant Chinese lines, is comprised of 771 recombinant inbred lines genotyped by sequencing at 43,320 single nucleotide polymorphisms. We phenotyped the NAM population for emergence, seedling vigor, and agronomic traits (>75,000 data points from ~16,000 plots) in multi-environment field trials in Kansas under natural chilling stress (sown 30–45 days early) and normal growing conditions. Joint linkage mapping with early-planted field phenotypes revealed an oligogenic architecture, with 5–10 chilling tolerance loci explaining 20–41% of variation. Surprisingly, several of the major chilling tolerance loci co-localize precisely with the classical grain tannin (*Tan1* and *Tan2*) and dwarfing genes (*Dw1* and *Dw3*) that were under strong directional selection in the US during the 20th century. These findings suggest that chilling sensitivity was inadvertently selected due to coinheritance with desired nontannin and dwarfing alleles. The characterization of genetic architecture with NAM reveals why past chilling tolerance breeding was stymied and provides a path for genomics-enabled breeding of chilling tolerance.

KEYWORDS
Multiparental population
Crop evolution
Climate adaptation
Cold tolerance
Antagonistic pleiotropy
Linkage drag

Adaptation to diverse environments has generated abundant genetic diversity in wild and domesticated plant species (Anderson *et al.* 2011; Meyer and Purugganan 2013). The genetic architecture of adaptation has been intensively studied both theoretically and empirically, but remains contentious. For instance, much debate surrounds the relative contributions of standing genetic variation *vs.* new mutation (Barrett and Schluter 2008), oligogenic *vs.* polygenic variation (Orr 2005), and pleiotropic *vs.* independent effects (Paaby and Rockman 2013). Despite the importance of adaptive variation in crop improvement, the genomic basis of local adaptation underlying abiotic stressors remains poorly understood (Olsen and Wendel 2013). Understanding the genomic basis of adaptation in crops can guide breeding strategies and facilitate transfer of adaptive traits for new climate-resilient varieties (Soyk *et al.* 2017; Zhu *et al.* 2018; Li *et al.* 2018).

Cold temperatures are a major factor limiting plant productivity globally for both wild plants and crops (Cramer *et al.* 1999). Tropical-origin crops (*e.g.*, maize, rice, tomato, cotton, sorghum) are typically sensitive to chilling temperatures (0–15°), which limits their range

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and/or growing season in temperate climates (Lyons 1973; Long and Spence 2013). Developing chilling-tolerant varieties could facilitate early planting to extend growing seasons, prevent soil moisture depletion, and shift growth and flowering to more favorable evapotranspirative conditions (Tuberosa 2012; Ma *et al.* 2015). For breeding chilling tolerance in tropical-origin crops, chilling-adapted germplasm from high-latitude zones and high-altitude tropical regions can be targeted as donors. Molecular mechanisms underlying cold tolerance (chilling and/or freezing temperatures) include C-repeat binding factor (CBF) regulon cold signaling (Thomashow 2001; Park *et al.* 2015; Wang *et al.* 2018), jasmonate signaling (Hu *et al.* 2013; Mao *et al.* 2019), and lipid remodelling (Li *et al.* 2004; Moellering *et al.* 2010).

Sorghum, a tropical-origin warm-season (C4) cereal, is among the major crops that are generally susceptible to chilling (Franks *et al.* 2006). Sorghum originated in tropical Africa (c. 5–10 thousand years ago) and diffused to temperate areas, including China (c. 800 years ago) and the United States (c. 200 years ago) (Kimber 2000). Diffusion of tropical sorghums to temperate climates has led to commercial sorghum industries covering several million hectares in US, Australia, Argentina, and China (Monk *et al.* 2014). Using a phytogeographic approach (Vavilov 1951), Chinese sorghum were targeted as chilling tolerance donors for conventional breeding starting in the 1960s (Stickler *et al.* 1962). However, characteristics of Chinese sorghums that are undesirable for US grain sorghum, particularly grain tannins and tall stature (>2 m) (Franks *et al.* 2006), have hampered breeding. Biparental linkage mapping identified chilling tolerance QTL tagged by the same molecular markers as grain tannins and plant height, but small populations and low marker density limited dissection of these traits (Knoll *et al.* 2008; Brown *et al.* 2008; Xiang 2009; Burow *et al.* 2010; Wu *et al.* 2012). Several classical tannin (*Tan1* and *Tan2*) and dwarfing (*Dw1–Dw4*) genes (Stephens 1946; Quinby and Karper 1954) have been cloned in recent years (Multani *et al.* 2003; Wu *et al.* 2012; Hilley *et al.* 2016, 2017), which could aid further trait dissection.

NAM populations can provide increased power for dissecting complex traits (Buckler *et al.* 2009; Ogut *et al.* 2015), particularly for adaptive traits where population structure confounds associations studies of natural populations (Bouchet *et al.* 2017). To dissect the genetic architecture of early-season chilling tolerance in sorghum, we developed and deployed a new nested association mapping (chilling NAM) resource. The chilling NAM population addresses a gap in existing sorghum NAM resources (Bouchet *et al.* 2017) by including contrasting temperate-adapted founders, three chilling-tolerant Chinese founders and the chilling-susceptible reference line BTx623 (Paterson *et al.* 2009). We used the chilling NAM to dissect the genetic architecture of sorghum early-season chilling tolerance at a high resolution based on natural field stress conditions. This NAM study provides insights into the origin and persistence of chilling sensitivity in US grain sorghum, and reveals new strategies for genomics-enabled breeding in this system.

MATERIALS AND METHODS

Population development

The chilling NAM population consists of three biparental populations that share a common parent, the US reference line BTx623 (Paterson *et al.* 2009) (Figure S1). The NAM founders were selected based on their contrasting chilling responses from early planting in preliminary studies in Lubbock, Texas. Chilling-sensitive BTx623 was used as the seed parent in crosses with three chilling-tolerant Chinese founders, Niu Sheng Zui (NSZ; PI 568016), Hong Ke Zi (HKZ; PI 567946), and Kaoliang (Kao; PI 562744) in Lubbock, Texas. BTx623 is derived from Combine Kafir × SC170, an Ethiopian zerazera caudatum (Menz *et al.* 2004). The resulting F_1 progenies were self-pollinated to generate three segregating F_2 populations. RILs were developed using single-seed descent by selfing to the F_6 generation in Lubbock, Texas (summer nursery) and Guayanilla, Puerto Rico (winter nursery). The $F_{6:7}$ RILs were derived by combining seeds of 3–4 uniform panicles. Additional seed increase of the NAM population was conducted in Puerto Vallarta, Mexico (winter nursery), by selfing the $F_{6:7}$ plants to derive $F_{6:8}$ RILs. Below, the Chinese founder name will be used when referring to a given RIL family (*e.g.*, the NSZ family).

Early- and normal-planted field trials

Six early- and two normal-planted field trials were conducted in 2016, 2017, and 2018 in Kansas (Table S1). Three locations, two in eastern Kansas [Ashland Bottoms (AB), 39.14N -96.63W; Manhattan (MN), 39.21N -96.60W] and one in western Kansas [Agricultural Research Center, Hays (HA), 38.86N -99.33W], were used for field trials (Figure 1A). Abbreviated location name and the last two digits of the year (*e.g.*, AB16 for Ashland Bottoms 2016) were assigned for each field trial. A suffix was added to the AB16 field trials, AB16_b1 and AB16_b2, as both were planted in AB with an interval of two weeks between plantings. The $F_{6:7}$ RILs were planted in AB16, while $F_{6:8}$ RILs were planted in AB, MN, and HA in 2017 and 2018. Each field trial contained two replicates of the NAM population. The NAM RILs were randomized within family in 2016, and completely randomized in 2017 and 2018 in each replicate block (Figures 1A and S2). Controls in each field trial comprised chilling-tolerant Chinese accessions NSZ, HKZ, Kao, and Shan Qui Red (SQR; PI 656025), chilling-sensitive inbreds BTx623 and RTx430, and US commercial grain sorghum hybrid Pioneer 84G62.

Five early-planted (EP, natural chilling stress) trials were sown in April and one in early May (MN17), 30–45 days earlier than normal sorghum planting in Kansas (Grain Sorghum Production Handbook 1998). The EP trials, except MN17, experienced chilling stress (<15°) during emergence (Table S1 and Figure S3). Optimal temperatures (>15°) prevailed in MN17 during emergence, but one-week-old seedlings experienced chilling stress (5–13°). Normal-planted (NP, optimal temperature) field trial was sown in June when the soil temperatures were optimal for sorghum cultivation (>15°). AB18 was considered as the second NP trial, although planted in early May, as optimal conditions prevailed during emergence and seedling growth.

Field phenotyping

Seedling phenotypes of the NAM population were evaluated under early- and normal-planted field trials. Prefixes EP and NP were included for each seedling trait to differentiate phenotypes from early- and normal-planted trials, respectively. Emergence count (EC) was scored on a scale of 1–5 that represented 20, 40, 60, 80, and 100% emergence, respectively. Three seedling vigor (SV) ratings (SV1–SV3), scored independently of emergence count, were collected at week-1, -2, and -4, respectively, after emergence. SV was scored on a rating scale of 1–5 with a rating of 1 and 5 for low and robust vigor, respectively (Figure 1B). A previously described SV scale (Maiti *et al.* 1981) was modified (1 for high and 5 for low SV) for consistency with EC rating. Repeatability of SV rating, SV2 (AB17) and SV3 (MN17), was tested with SV ratings collected simultaneously by different individuals. Early-planted damage rating (EPDR), based on visual damage observed two days after a severe chilling stress event, was scored on a 1–5 rating scale representing seedling death/severe leaf-tip burning, leaf-tip burning, severe chlorosis, mild/partial chlorosis, and no chilling damage

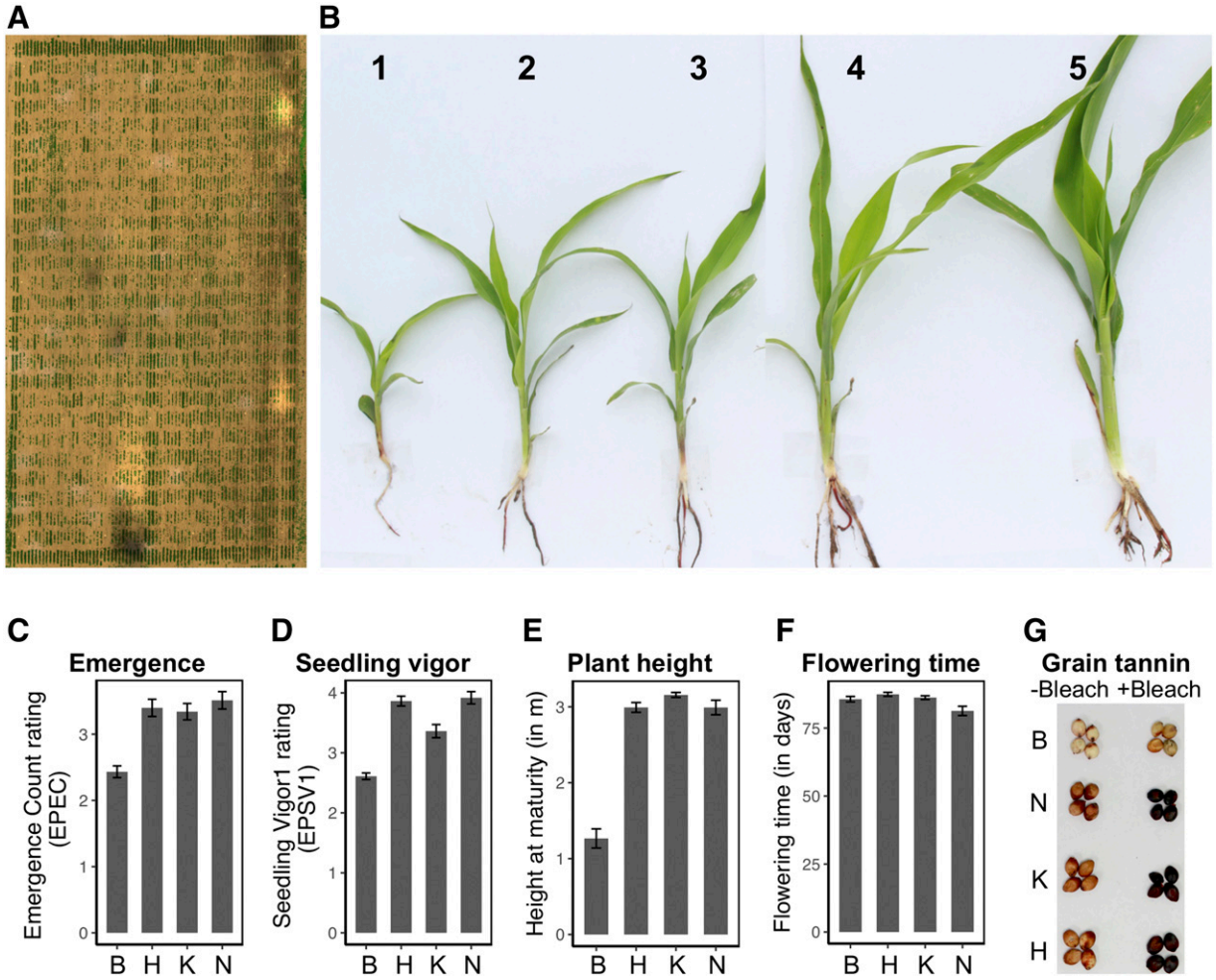

Figure 1 Chinese sorghums harbor early-season chilling tolerance and characteristics undesirable for US grain sorghums. (A) Aerial image of an early-planted field (AB17) trial for chilling tolerance phenotyping based on stitched RGB imagery (B) Seedling vigor rating used in field trials. In early-planted field trials, differences were observed in (C) emergence and (D) seedling vigor between the four NAM founders, B (BTx623), K (Kao), H (HKZ), and N (NSZ). Additionally, (E) significant variation in plant height at maturity, (F) no significant difference in flowering time (days after emergence), and (G) presence/absence variation in grain tannins were observed.

symptoms, respectively. Seedling height was measured manually one month after initial emergence in each location.

Plant height and flowering time (days to flowering after emergence), the major agronomic traits, were collected from three (AB16_b1, MN17, and MN18) and two (AB16_b1 and MN17) field trials, respectively. Agronomic suitability of the NAM population as US grain sorghum, which included semi-dwarf stature, panicle exertion, standability, and compact panicle architecture, was screened in AB16_b1. Presence or absence of grain tannins in field-grown samples (from Puerto Vallarta, Mexico) of each RIL was determined using bleach test with SQR, a Chinese accession containing grain tannin, as a positive control (Wu *et al.* 2012). Fifteen seeds from each RIL were transferred into a 2 ml tube and 1 ml bleach solution (3.5% sodium hypochlorite and 5% sodium hydroxide) was added. RILs with tannins turned black after 30 min and were scored as 1. By contrast, nontannin RIL seed did not change their color and were scored as 0.

Statistical analysis of phenotypes

Trait correlation between locations was determined using the averaged seedling trait ratings of two replicates from each field trial. Pearson pairwise correlation analysis was performed using *pairs.panels* function in psych R package. Broad sense heritability (H^2) estimate of EP and NP field phenotypes was calculated with seedling ratings from six EP and two NP field trials, respectively. Seedling traits H^2 was calculated from variance components generated with the *lme4* (Bates *et al.* 2015) R package as described earlier (Boyles *et al.* 2017). All components were treated as random effects and replicates were nested in location-by-year interaction:

$$lmer(Trait = (1\,|\,G) + (1\,|\,L) + (1\,|\,Y) + (1\,|\,R\%in\%L{:}Y) + (1\,|\,G{:}L) + (1\,|\,G{:}Y)$$

and broad-sense heritability was calculated using the equation:

$$H^2 = \frac{\sigma^2 G}{\sigma^2 G + \left(\frac{\sigma^2 G_{GxL}}{L}\right) + \left(\frac{\sigma^2 G_{GxY}}{Y}\right) + \left(\frac{\sigma^2 E}{LY}\right)}$$

where G is the genotype, L is the location, Y is the year, R is the replicate, and E is the error term. Environment main effects were not included in the denominator as they do not influence response

to selection (Holland *et al.* 2003). Best linear unbiased predictions (BLUPs) of EP and NP seedling traits were generated using the model for estimating H^2.

DNA extraction and genotyping

Genotyping-by-sequencing (GBS) was conducted on one-week-old seedlings of the $F_{6:7}$ RILs and the four NAM founders (Figure S1). Leaf tissue (~50 mg, pooled from three seedlings) from each RIL was transferred into a 96-deepwell plate, lyophilized, and stored at -80°. One ball bearing was added to each well and the leaf tissue was ground with a Retsch Mixer Mill MM400 tissue grinder (Vernon Hills, IL, USA). Genomic DNA was extracted using QIAGEN BioSprint 96 DNA plant kit (Germantown, MD, USA). DNA was quantified with Quant-iT PicoGreen dsDNA assay kit (Thermo Fisher Scientific, Grand Island, NY, USA) using Agilent 2100 Bioanalyzer (Santa Clara, CA, USA) at the Kansas State University Integrated Genomics Facility. Each sample was normalized to contain 10 ng/μl DNA using QIAgility Liquid Handling System (Germantown, MD, USA). Six μl of DNA was transferred to a 96-well PCR plate and adapters were added. *Ape*KI enzyme was used for restriction digestion and GBS libraries were prepared as described previously (Elshire *et al.* 2011; Morris *et al.* 2013a). Illumina HiSeq 2500 Rapid v2 sequencing system was used for 100-cycle single-end sequencing of two 384-multiplexed libraries at the University of Kansas Medical Center Genome Sequencing Facility.

GBS data from the chilling NAM resource was combined with previously published *Ape*KI GBS data from ~10,323 diverse accessions (Hu *et al.* 2019), aligned to the BTx623 reference genome v3.1 (McCormick *et al.* 2018), and SNP calling was performed using Tassel 5.0 GBS v2 pipeline (Glaubitz *et al.* 2014). GBS of the chilling NAM population provided 528,065 single nucleotide polymorphisms (SNPs) (Figure S1). GBS data were filtered for 80% missingness (PM) and 0.05 minor allele frequency (MAF, based on the chilling NAM population that contains 50% BTx623 alleles and ~16% alleles from each Chinese parent). 61,428 SNPs were retained and these SNPs were separated by individual chromosomes and imputed using Beagle 4.1 (Browning and Browning 2013). Additional filtering for markers and RILs with >15% residual heterozygosity retained 43,320 SNPs and 750 RILs for joint linkage mapping.

Population genetic analyses

Genetic structure of the chilling NAM population was characterized with respect to global sorghum germplasm. First, the chilling NAM and global accessions GBS data were filtered for 80 PM and 0.01 MAF (to avoid discarding rare alleles), and the retained SNPs (265K) were imputed using Beagle 4.1 (Browning and Browning 2013). Next, two PCA axes were built with previously published *Ape*KI GBS data of 401 global sorghum accessions (Morris *et al.* 2013a), and chilling NAM founders and RILs were projected on these axes. Principal component analysis (PCA) of global germplasm was performed using *prcomp* function in R. Coordinates for the chilling NAM population were calculated with the *predict* function in R.

Neighbor-joining analysis, using TASSEL 5.0 (Glaubitz *et al.* 2014), was conducted with 61,428 SNPs to characterize the genetic relatedness of the chilling NAM population. In TASSEL, modified Euclidean distance was used as distance model and neighbor-joining as the tree algorithm. A neighbor-joining tree was constructed with Ape package (Paradis *et al.* 2004) in R (R Core Team 2018). SNP density was calculated with VCFtools (Danecek *et al.* 2011) in 200kb windows. Linkage disequilibrium (LD) decay was estimated, using pairwise comparisons of ~55–70K GBS SNPs, individually for the three NAM families with PopLDdecay v.3.29 package (Zhang *et al.* 2018). LD decay of 176 Ethiopian and 29 Chinese landraces (genotyped previously with *Ape*KI) (Lasky *et al.* 2015) was estimated for comparison. Ethiopian and Chinese germplasm LD decay was calculated using ~100K and ~57K SNPs, respectively. Parameters were set for -MaxDist as 500 kb and -MAF as 0.05. LD decay curves were plotted based on r^2 and the distance between pairs of SNPs.

Linkage mapping analysis

The NAM founders genotypes were used for constructing genetic linkage maps with the R/qtl package (Broman *et al.* 2003). The level of heterozygosity between the parental lines was determined using VCFtools–het function (Danecek *et al.* 2011). The NAM founders were filtered for 20 PM and <0.4 MAF and the retained SNPs were used to retrieve the NAM population genotypes from the GBS dataset. SNP imputation was conducted for each family separately using Beagle 4.1 (Browning and Browning 2016). RILs with >85% missing data or >80 crossovers were dropped. Duplicate markers (*i.e.*, mapping to the same location; ~200 from each family) were identified using the *findDupMarkers* function and dropped with *drop.markers* based on the R/qtl tutorial recommendation. Genetic linkage maps for each NAM family were generated using the *Haldane* function. The *Droponemarker* function in R/qtl was used to discard problematic markers that increase chromosome length. Genetic linkage maps were reconstructed for each NAM family. Composite interval mapping (CIM) (Zeng 1994), with R/qtl, was used for performing linkage mapping and significant QTL were determined based on the threshold level defined by computing 1000 permutations. Allelic effects were defined as positive or negative effects of the BTx623 allele. LOD support interval for individual QTL was obtained with the *lodint* R/qtl function. CIM was performed with plant height, flowering time, and grain tannin data to validate the generated genetic linkage maps. BLUPs of seedling traits, EC and SV1–3, from early- and normal-planted field trials were used for CIM. Additionally, linkage mapping was performed for individual field trials with the averaged data of two replicates from each location.

Joint linkage mapping

Joint linkage mapping (JLM) was conducted with 43,320 GBS SNPs and seedling trait BLUPs from 750 RILs. In addition, JLM was performed individually for each location with the averaged data of two replicates. Mapping power and resolution of the chilling NAM population was validated using plant height, flowering time, and grain tannin data. Stepwise regression approach in TASSEL 5.0 (Glaubitz *et al.* 2014), which uses forward inclusion and backward elimination stepwise method, was used to perform JLM. Entry and exit limit of the forward and backward stepwise regressions was 0.001 and threshold cut off was set based on 1000 permutations. JLM was performed using the following equation:

$$y = b_o + \alpha_f u_f + \sum_{i=1}^{k} x_i b_i + e_i$$

where b_0 is the intercept, u_f is the effect of the family of founder line f obtained in the cross with the common parent (BTx623), α_f is the coefficient matrix relating u_f to y, b_i is the effect of the ith identified locus in the model, x_i is the incidence vector that relates b_i to y and k is the number of significant QTL in the final model. Allelic effect for each QTL was expressed relative to the BTx623 allele, where alleles with positive- or negative-additive effects were derived from BTx623 or Chinese founders, respectively. Based on the average genome-wide recombination rate of 2.0 cM/Mb for sorghum

(Mace *et al.* 2009; Bouchet *et al.* 2017), QTL for one or more seedling traits that mapped within a 2 Mb interval were assigned a common name. For example, *qSbCT04.62* to describe QTL detected on chromosome 4 close to 62 Mb.

Sequence variant analysis

CBF and *Tan1* genes, colocalizing with chilling tolerance QTL, were used for sequence variant analysis. Two overlapping primer pairs were used to amplify these genes from the Chinese founders (primer sequences are included in Table S2). 50% glycerol and 25mM $MgCl_2$ were added to the master mix for stabilizing the PCR reaction. PCR product purification and Sanger sequencing were performed at GENEWIZ (South Plainfield, NJ). Clustal Omega and Expasy translate were used for sequence alignment and predicting the peptide sequences of *CBF1* and *Tan1* genes.

Ecophysiological crop modeling

CERES-Sorghum crop model (White *et al.* 2015) in the Decision Support Systems for Agro-technology Transfer-Crop Simulation Model software (Jones *et al.* 2003) was used to predict the value of early planting for grain sorghum in the Kansas production environment. This model simulates daily physiological processes using a base temperature of 8° (White *et al.* 2015) and has effectively predicted sorghum grain yield in Kansas (Staggenborg and Vanderlip 2005; Araya *et al.* 2018). We consider that this model assumes chilling tolerance by default, since it does not model damage due to chilling temperatures. A full-season (late-maturing) photoperiod insensitive grain sorghum hybrid, used in previous crop modeling, was used in this study (Araya *et al.* 2018). Simulations were performed under rainfed conditions at four representative Kansas locations, Colby (39.39N, -101.06W), Garden City (37.99N, -101.81W), Hays (38.84N, -99.34W), and Manhattan (39.20N, -96.55W), from a 30 year period (1986–2015). Historical weather data for each of these locations was obtained from Kansas Mesonet (2019). Simulations were started on January 1 to account for the effect of precipitation on soil moisture and the onset of soil evaporation. Early (April 15), normal (May 15), and late (June 15) planting scenarios were simulated and (i) available precipitation, (ii) days of water stress after anthesis, and (iii) final grain yield were analyzed.

Data availability

Sequencing data are available in the NCBI Sequence Read Archive under project accession SRP8838986. Figure S1 depicts the experimental design for mapping early-season chilling tolerance. Figure S2 contains the field layout of trials. Figure S3 shows daily minimum air temperature across field trials. Figure S4 contains seedling traits of US and Chinese founders. Figure S5 includes SNP density map and genetic structure of the chilling NAM population. Figures S6 and S7 show trait correlation across trials. Figure S8 visualizes genetic linkage maps for each family. Figure S9–S12 are comparisons of EP and NP JLM for emergence count (EC), seedling vigor (SV1, SV2, and SV3). Figures S13–18 are JLM from individual field trials. Figures S19 and S20 have CIM of plant height and grain tannins, respectively. Figure S21 has JLM of plant height and flowering time, and Figure S22 has JLM with grain tannins. Figure S23 has JLM QTL for chilling tolerance in tannin *vs.* nontannin RILs. Figure S24 describes ecophysiological crop simulations of early-season chilling tolerance. File S1 contains genetic linkage maps. File S2 contains Hapmap data. File S3 contains R scripts used for analyses and seedling trait BLUPs. Table S1 has phenotyping details for each field trial. Table S2 has primers used to amplify *CBF* and *Tan1* genes. Tables S3 and S4 contain CIM QTL from EP- and NP seedling trait BLUPs, respectively. Tables S5 and S6 have JLM QTL from NP seedling trait BLUPs and agronomic traits, respectively. Table S7 is a comparison of CIM QTL with previously-mapped chilling tolerance QTL. Table S8 has EP seedling trait ratings of RILs selected for favorable plant height. The chilling NAM population seeds will be submitted to the USDA National Plant Germplasm System(https://www.ars-grin.gov/). Please contact G.B. (gloria.burow@ars.usda.gov) or the corresponding author for availability. Supplemental material available at figshare: https://doi.org/10.25387/g3.9755336.

RESULTS

Development of NAM population for chilling tolerance studies

The chilling NAM population was generated from crosses of a US reference line BTx623 with three Chinese lines, NSZ, Kao, and HKZ (Figure S1). The resulting chilling NAM population ($n = 771$) comprised 293, 256, and 222 RILs for the NSZ, Kao, and HKZ families, respectively. Our chilling tolerance studies of the NAM founders and RILs were based on natural chilling events in field trials sown 30-45 days earlier than normal. In early-planted field trial (Figures 1A–B) the Chinese founders had significantly greater emergence and seedling vigor ($P < 0.05$) than BTx623 (Figures 1C, 1D, and S4). By contrast, no difference was observed in emergence (NPEC) between the founders under normal-planting (Figure S4). Chinese founder lines were much taller (~3 m) at maturity than BTx623 (1.2 m) (Figure 1E), but little variation was observed for flowering time among the founder lines (4–5 d; $P < 0.05$; Figure 1F). Grain tannins were present in the Chinese accessions and absent in BTx623 (Figure 1G).

Genetic properties of the chilling NAM population

The filtered GBS data set for the chilling NAM population comprised genotypes at 43,320 SNPs. SNP densities were higher in telomeres than pericentromeric regions (Figure S5A). To check the population's quality and understand its genetic structure, NAM RILs and founders were projected onto PCA axes built from a global sorghum diversity panel (Figure 2A), which reflect geographic origin and botanical race (Harlan and de Wet 1972; Morris *et al.* 2013a). As expected, the Chinese founders clustered with durra sorghums of Asia and East Africa, while BTx623 was positioned midway between kafir and caudatum clusters, consistent with its pedigree (Menz *et al.* 2004) (Figure 2A). The three half-sib families of the chilling NAM population were clustered together, midway between the Chinese founders and BTx623. NJ analysis (Figure S5B) and PCA (Figure S5C) of the chilling NAM population by itself confirmed the expected family structure for NSZ and Kao, with each family forming a single cluster. Two clusters were observed for the HKZ family. We assigned HKZ RILs into HKZa ($n_{RIL} = 121$) or HKZb ($n_{RIL} = 101$) subfamilies, with the HKZb subfamily representing the cluster with PC1 > 40 (and the longer branch on NJ dendrogram). The LD rate decay (to genome-wide background) was slower in NAM families (~500 kb) compared to diverse accessions from China and Ethiopia (~20 kb) (Figure 2B).

Repeatability and heritability of field phenotypes

RILs were scored for emergence and seedling vigor under early- and normal-planted field trials. Early (EPSV1) and later (EPSV2, EPSV3) seedling vigor ratings were strongly correlated (0.7–0.8), as were ratings made by different individuals on the same day (0.7–0.8) (Figure S6). By contrast, the correlation across RILs between early- and normal-planted seedling traits was low (0.1–0.3). Broad sense heritability (H^2) across locations and years for early-planted seedling traits was intermediate (0.4–0.5) (Table 1), while H^2 was higher (0.5–0.8) for

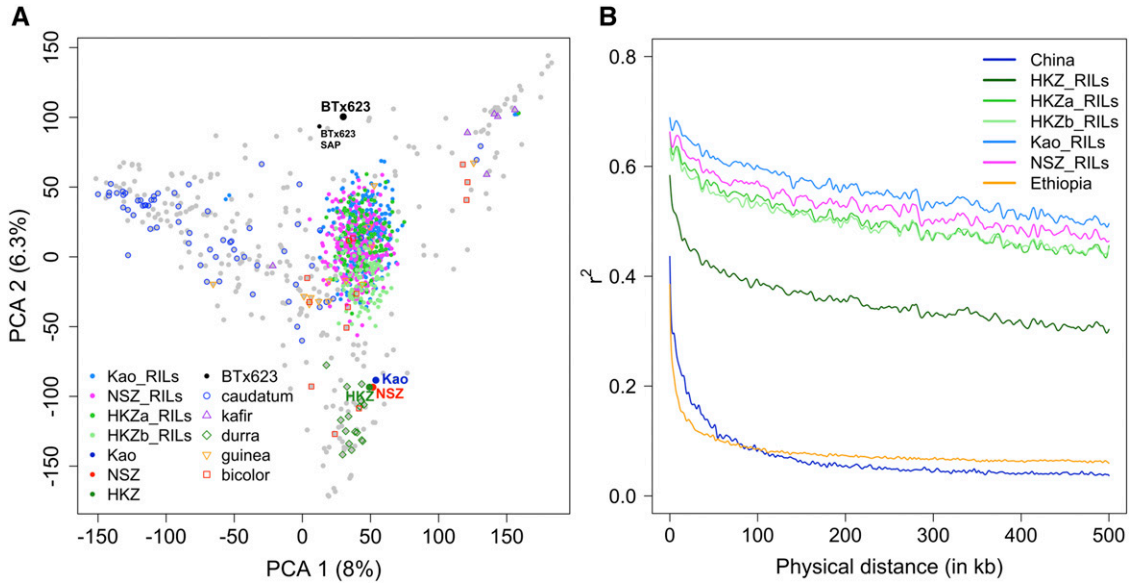

Figure 2 Genetic properties of the chilling NAM population (A) Principal component analysis (PCA) of the NAM (n_{RIL} = 771) plotted on PCA axes built with 401 accessions of the global sorghum diversity germplasm. Major botanical races (Caudatum, Kafir, Durra, Guinea, and Bicolor) of global accessions are noted with symbols (B) Linkage disequilibrium (LD) decay of the NSZ, HKZa, HKZb, and Kao families. LD decay rate of diverse accessions from China (n = 29) and Ethiopia (n = 176) are presented for comparison.

seedling traits from normal-planted field trials. H^2 for seedling height (in early-planted field trials) was close to zero (0.03), while plant height at maturity was highly heritable (0.9). Based on the averaged data of two replicates within each field trial, low to intermediate correlation (0.1–0.4) was observed with the same seedling trait among locations for early-planted trials (Figure S7).

Composite interval mapping of early-season chilling tolerance

Genetic linkage maps were constructed for each family (NSZ: 1341 markers, 257 RILs; Kao: 1043 markers, 219 RILs; HKZa: 1150 markers, 107 RILs) (Figure S8). Map lengths were similar for the NSZ, Kao, and HKZa families (1403 cM, 1381 cM, and 1295 cM, respectively) and individual RILs contained 2–4 crossovers. To map putative chilling tolerance loci, composite interval mapping (CIM) was first conducted in individual families using ~1000–1300 markers and early-planted seedling trait BLUPs (EPEC, EPSV1–3). CIM detected 6–8 QTL, which explained 16–28%, 8–23%, and 12–36% of variation for early-planted seedling traits in the HKZa, Kao, and NSZ families, respectively (Table S3). The QTL on chromosome 4 was detected in all NAM families, with the positive allele inherited from the Chinese founder in each case. CIM of normal-planted seedling BLUPs (NPEC and NPSV1–NPSV3) identified 4–9 QTL contributing to emergence and SV in the HKZa, Kao, and NSZ families, respectively. Few overlaps were observed among QTL detected for early- and normal-planted seedling traits (Tables S3 and S4). As chilling stress varied among locations (Figure S3), QTL mapping was conducted for each field trial separately to check the stability of QTL across locations. The QTL on chromosomes 4 and 7 were detected across families in four and two early-planted trials, respectively.

Joint linkage mapping of early-season chilling tolerance

To leverage data across families, JLM was performed with 43,320 SNPs and field phenotypes from 750 RILs (including the HKZb family) (Figure 3A–E). JLM of seedling trait BLUPs (derived from ~12,000 early-planted plots) identified 15 QTL, seven of which were detected for multiple seedling traits (Figure 3D and Table 2). Each QTL explained 1–9% of phenotypic variation. In total, the QTL explained 21–41% variation for emergence and seedling vigor. Positive alleles were inherited from the Chinese founders, except for the allele at chromosome 3. The QTL on chromosomes 2 and 4 were detected for every early-planted seedling trait. The chromosome 1 and 5 QTL were detected with all seedling vigor traits, while chromosome 7 and 9 were mapped with two early-planted seedling traits (Figure 3D). The QTL on chromosomes 2 and 4 colocalized (<1 Mb) with classical tannin genes, *Tan2* and *Tan1* (Wu *et al.* 2012; Morris *et al.* 2013b), and chromosomes 7 and 9 loci colocalized with classical dwarfing genes, *Dw3* and *Dw1* (Multani *et al.* 2003; Hilley *et al.* 2016). JLM of normal-planted traits mapped different QTL for emergence, but few overlapped with QTL for early-planted seedling vigor (Figures 3C and S9–S12, and Table S5).

To check the stability of QTL across locations and years, JLM was performed separately by location. The QTL on chromosome 9 was detected in three early-planted locations, while QTL on chromosomes 2 and 7 were mapped in two locations (Figure 3B and S13–S18). The chromosome 4 QTL was consistently detected across early-planted field

Table 1 Broad-sense heritability (H^2) of early- and normal-planted field traits

Seedling traits	H^2 early planting[a]	H^2 normal planting[b]
Emergence count (EC)	0.45	0.53
Seedling vigor1 (SV1)	0.52	0.57
Seedling vigor2 (SV2)	0.39	0.78
Seedling vigor3 (SV3)	0.37	0.53
Damage rating (DR)	0.35	—
Seedling height	0.03	—
Plant height at maturity	0.93	—

[a]Field phenotypes from six early-planted trials.
[b]Field phenotypes from two normal-planted trials.

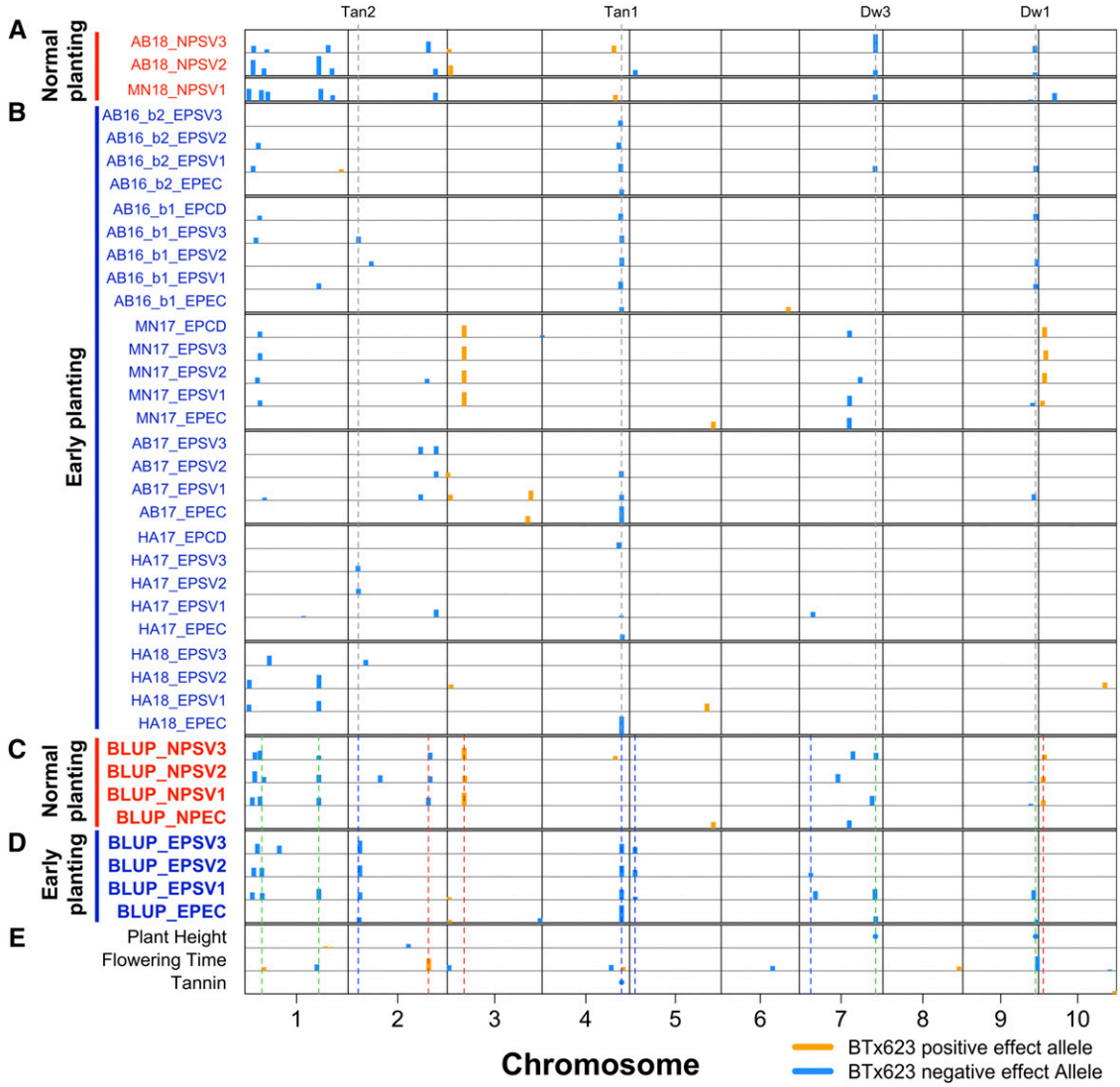

Figure 3 Joint linkage mapping (JLM) of chilling tolerance and undesirable traits JLM of seedling traits from individual (A) normal (NP) and (B) early planting (EP) field trials. Field location and year were included as prefixes for each seedling trait. Five NP traits that failed to detect QTL were excluded from the figure but used for calculating NP seedling trait BLUPs. JLM with seedling trait BLUPs, generated with ~75,000 data points from ~16,000 field plots, from (C) normal and (D) early planting. Additionally, (E) JLM of plant height, flowering time, and grain tannins were included. Classical dwarfing and tannin genes were noted with gray dashed lines. Chilling tolerance QTL detected under early planting are noted with blue dashed lines and green lines noted chilling tolerance QTL detected under both early and normal planting. The QTL under normal planting were noted with red dashed lines. Positive or negative effects of the BTx623 allele was indicated in orange or blue colors, respectively. The percentage of variation explained is proportional to the width of the box for each locus and loci explaining phenotypic variation >10% are noted with circles. Abbreviations: EC, emergence count; SV1–3, seedling vigor1–3; CD, chilling damage.

locations and years. The only exception was the MN17 field trial, which emerged under optimal conditions and experienced chilling one week later, where the chromosome 4 QTL was not detected (Figures 3B and S16). Among the loci detected with JLM of field phenotypes from early- and normal-planted individual field trials (Figures 3A–B), few overlaps were observed.

The most significant and consistent QTL (*qSbCT04.62*; Table 2) colocalized with CBF gene Sobic.004G283201 (120 kb from the peak SNP), ortholog of the canonical Arabidopsis cold acclimation regulator *CBF1* (Thomashow 2001; Park *et al.* 2015). However, sequencing of the CBF gene from the Chinese founders revealed no change in their predicted peptide relative to the BTx623 reference sequence.

Mapping for agronomic traits and grain tannin

CIM and JLM was conducted to identify loci underlying plant height, flowering time, and grain tannins. CIM detected three plant height QTL in the HKZa family (Table S6 and Figure S19), and two each in the NSZ and Kao families, explaining 30–82% of plant height variation. Two plant height QTL, detected on chromosomes 7 and 9, colocalized with classical dwarfing genes *Dw3* and *Dw1*, respectively (Multani *et al.* 2003; Hilley *et al.* 2016). JLM identified six plant height QTL, of which alleles at four and two QTL contained negative and positive effects, respectively (Figures 3C and S21, and Table 3). Three QTL of major effect explained 85% plant height variation. Major height loci were 12 kb and 0.1 Mb from *Dw3* and *Dw1* genes, respectively.

Table 2 Joint linkage mapping (JLM) with early-planted field phenotypes

Trait[a]	QTL	QTL_SNP	PVE[b]	Additive effect[c]	Known loci[d]	Distance to known loci	QTL name[e]
EPEC	*qSbEPEC_4-62*	S4_62368531	9.2	−0.08	*Tan1*	53 kb	*qSbCT04.62*
	qSbEPEC_3-72	S3_72791601	2.4	0.01			
	qSbEPEC_2-08	S2_8672301	2.8	−0.06	*Tan2*	0.6 Mb	*qSbCT02.08*
	qSbEPEC_7-59	S7_59915577	3.3	−0.08	*Dw3*	93 kb	*qSbCT07.59*
	qSbEPEC_9-58	S9_58070153	1.5	−0.04	*Dw1*	1 Mb	*qSbCT09.57*
	qSbEPEC_3-01	S3_1779472	1.4	0.03			
EPSV1	*qSbEPSV1_4-62*	S4_62368531	5.4	−0.07	*Tan1*	53 kb	*qSbCT04.62*
	qSbEPSV1_9-55	S9_55625332	5	−0.07	*Dw1*	1.4 Mb	*qSbCT09.57*
	qSbEPSV1_1-57	S1_57941435	5.7	−0.11			*qSbCT01.57*
	qSbEPSV1_1-05	S1_5730743	3.9	−0.06			*qSbCT01.06*
	qSbEPSV1_7-12	S7_12580350	4.5	−0.06			*qSbCT07.10*
	qSbEPSV1_2-09	S2_9260382	3.9	−0.06	*Tan2*	1.2 Mb	*qSbCT02.08*
	qSbEPSV1_3-01	S3_1447612	1.4	0.03			
	qSbEPSV1_5-04	S5_4403613	1.6	−0.04			*qSbCT05.04*
	qSbEPSV1_1-13	S1_13526795	3.5	−0.1			*qSbCT01.13*
	qSbEPSV1_7-59	S7_59290017	5.6	−0.08	*Dw3*	0.5 Mb	*qSbCT07.59*
EPSV2	*qSbEPSV2_4-62*	S4_62455479	5.8	−0.05	*Tan1*	0.1 Mb	*qSbCT04.62*
	qSbEPSV2_2-09	S2_9218398	6	−0.05	*Tan2*	1.2 Mb	*qSbCT02.08*
	qSbEPSV2_5-04	S5_4284787	3.6	−0.04			*qSbCT05.04*
	qSbEPSV2_1-13	S1_13188261	4.6	−0.06			*qSbCT01.13*
	qSbEPSV2_1-06	S1_6902771	4.8	−0.05			*qSbCT01.06*
	qSbEPSV2_7-08	S7_8916696	2.1	−0.05			*qSbCT07.10*
EPSV3	*qSbEPSV3_2-09*	S2_9218398	6.8	−0.05	*Tan2*	1.2 Mb	*qSbCT02.08*
	qSbEPSV3_4-62	S4_62455479	5.2	−0.05	*Tan1*	0.1 Mb	*qSbCT04.62*
	qSbEPSV3_1-09	S1_9756192	5.2	−0.06			*qSbCT01.13*
	qSbEPSV3_1-26	S1_26930469	4.3	−0.05			
	qSbEPSV3_5-04	S5_4284787	3.5	−0.03			*qSbCT05.04*

[a] Early-planted emergence count (EPEC) and seedling vigor (EPSV1–3) BLUPS were used for JLM.
[b] Percentage of variation explained (PVE).
[c] Positive or negative effects of the BTx623 allele.
[d] Previously characterized genes colocalizing with the mapped QTL.
[e] QTL in 2 Mb interval, detected with different seedling traits, were assigned a common name.

Although flowering time varied little among the founders (Figure 1E), transgressive segregation enabled detection of seven flowering time loci (four, two, and one QTL in the NSZ, Kao, and HKZa families, respectively) which explain 20–28% of variation (Table S6). JLM with flowering time detected 10 QTL that explained 33% variation (Figures 3C and S21, and Table 3), three of which co-localized with previously identified flowering time/maturity genes, *TOC1*/*CN2*, *ma1*, and *CN8*. CIM of grain tannin presence/absence identified a major QTL on chromosome 4 in each family, with the Chinese parent conferring tannin presence allele in each case (Figure S20). The locus colocalizing with *Tan1* explained 77, 34, and 100% of grain tannin variation in the HKZa, NSZ, and Kao families, respectively (Table S6). JLM identified two tannin loci, one mapped ∼70 kb from *Tan1* and the other mapped ∼1.4 Mb from an earlier reported *Tan2* candidate gene (Wu *et al.* 2012; Morris *et al.* 2013b) (Figure S22, and Table 3).

DISCUSSION

A NAM resource to dissect the genetic architecture of chilling tolerance

Characterizing the genetic architecture of adaptive traits provides insight into mechanisms of adaptation (Orr 2005) and guides strategies for breeding (Bernardo 2008). The NAM approach has been used to increase power and accuracy for dissection of complex adaptive traits in several widely adapted crop species (Buckler *et al.* 2009; Nice *et al.* 2016; Bouchet *et al.* 2017). By using temperate-adapted founders with contrasting chilling responses (Figures 1C, 1D, and S4), the chilling NAM resource addresses a gap in available sorghum NAM resources (Bouchet *et al.* 2017). Together, the chilling NAM and global NAM population (Bouchet *et al.* 2017) make up a resource of >3000 lines for complex trait dissection in sorghum. Given the founder lines originated from different botanical races (kafir-caudatum *vs.* durra; Figure 2A), the chilling NAM population should harbor abundant diversity for future studies of adaptive traits. Anecdotal field observations suggest the population harbors variation in vegetative pigmentation, disease susceptibility, and panicle and stem architecture.

The quality of the chilling NAM resource (*i.e.*, RILs and corresponding SNP genotypes) developed in our study is validated by the precise mapping (<100 kb) of cloned dwarfing (*Dw1* and *Dw3*) and tannin (*Tan1*) genes (Figure 3, Table 3). Similarly, several major QTL (*qSbCT04.62*, *qSbCT02.08*, *qSbCT07.59*, and *qSbCT09.57*) were encompassed within the QTL intervals detected previously (Knoll *et al.* 2008; Burow *et al.* 2010) (Table S7). Notably, however, the greater population size (∼4–fivefold) and marker density (>100-fold) with NAM relative to earlier studies greatly improved the mapping resolution (>10-fold; Table S7) and power (*i.e.*, several additional loci identified). Family structure and LD decay of the chilling NAM population generally matches expectations based on population design and observations from previous NAM populations (Bouchet *et al.* 2017). Genotypic (Figure 2A) and phenotypic similarity of HKZa and HKZb RILs suggest that the differentiation is due to residual heterozygosity in the HKZ founder or pollen contamination from another Chinese accession. Given that inbreeding coefficient (*F*) is similar between Chinese parents (HKZ 0.8; Kao 0.9; NSZ 0.76; and BTx623 0.92) and parent-unique SNPs are absent only in the HKZb family, the HKZb RILs were most likely derived from pollen contamination. However, uncertainty

Table 3 Joint linkage mapping of plant height, flowering time, and grain tannins

Trait[a]	QTL	QTL_SNP	PVE[b]	Additive effect[c]	Known loci[d]	Distance to known loci
PHT	*qSbPHT_7-59*	S7_59675001	32	−21	*Dw3*	0.1 Mb
	qSbPHT_9-57	S9_57051085	20	−17	*Dw1*	12 kb
	qSbPHT_1-67	S1_67896587	0.5	0.7		
	qSbPHT_2-47	S2_47294140	2	−5		
	qSbPHT_7-59	S7_59956049	33	−21	*Dw3*	0.1 Mb
	qSbPHT_1-63	S1_63253487	1	7		
FT	*qSbFT_9-58*	S9_58468998	8	−1.5	*CN8*	3.5 Mb
	qSbFT_2-64	S2_63261883	6	1.4		
	qSbFT_8-59	S8_59740114	2	0.9		
	qSbFT_1-56	S1_56436041	3	−1		
	qSbFT_3-01	S3_1441099	3.03	−0.96		
	qSbFT_4-63	S4_63556402	2.01	0.82		
	qSbFT_4-54	S4_54231126	3.11	−1.32	*CN2*	8.6 Mb
	qSbFT_1-14	S1_14862315	1.97	0.92		
	qSbFT_10-56	S10_56045853	0.69	−0.4		
	qSbFT_6-40	S6_40299229	2.55	−0.89	*Ma1*	5 kb
Tannin	*qSbTan_4-62*	S4_62389178	72	−0.4	*Tan1*	73 kb
	qSbTan_4-62	S4_62261292	46	−0.4	*Tan1*	54 kb
	qSbTan_4-61	S4_61963287	22	−0.3	*Tan1*	0.3 Mb
	qSbTan_2-09	S2_9390193	0.06	0.02	*Tan2*	1.4 Mb
	qSbTan_10-59	S10_59593345	2	0.2		

[a] Plant height (PHT), flowering time (FT), and grain tannin phenotypes were used for JLM.
[b] Percentage of variation explained (PVE).
[c] Positive or negative effects of the BTx623 allele.
[d] Previously characterized PHT, FT, and grain tannin genes colocalizing with the mapped QTL.

regarding the pedigree of HKZb RILs does not diminish their usefulness as a part of the NAM resource (*e.g.*, Figure 3).

QTL mapping from multi-environment trials clearly identified a major oligogenic component of chilling tolerance (Figure 3), consistent with previous work (Knoll *et al.* 2008; Burow *et al.* 2010; Fiedler *et al.* 2016; Ortiz *et al.* 2017). In keeping with the breeding goals, we considered all QTL that controlled performance under chilling stress (emergence, seedling vigor, or both) as chilling tolerance loci (Table 2), regardless of whether they also controlled performance under normal conditions. As chilling tolerance trials were conducted in a field environment, heritability and QTL effect sizes (Tables 1 and 2) were somewhat reduced compared to previous experiments under controlled conditions (Knoll *et al.* 2008). While replicability of field phenotyping for abiotic stress is a major challenge (Araus and Cairns 2014), observing plant performance under field conditions may increase the likelihood that genetic discoveries will translate to farmer fields (Cobb *et al.* 2018). A common limitation for molecular breeding of stress tolerance has been a lack of QTL stability (*i.e.*, QTL × environment interaction) (Bernardo 2008). The overlapping of multi-environment chilling tolerance QTL from this study with QTL previously identified in the fields in Texas and Indiana (Table S7) provides evidence of their stability across a wide range of early-season chilling scenarios.

The genetic basis of early-season chilling tolerance

Molecular networks for cold sensing and response appear to be largely conserved across plants (Knight and Knight 2012; Dong *et al.* 2019). These findings are consistent with long-standing observations of homologous variation in cold tolerance across diverse grasses, including sorghum (Vavilov 1951). For this reason, we considered whether NAM provides evidence that chilling tolerance in Chinese sorghum is due to derived variation at canonical cold tolerance genes (*e.g.*, *CBF*s, *COLD1*, *SENSITIVE TO FREEZING2*, etc). Overall, we found little evidence that the chilling tolerance in Chinese sorghum is due to variation in canonical cold regulators (*i.e.*, little localization between QTL and sorghum orthologs of known plant cold tolerance genes). For instance, the *CBF* gene near the chilling tolerance QTL on chromosome 4 shows no coding sequence differences among the founder lines and a previous study showed no chilling-responsive expression of this *CBF* in chilling-tolerant NSZ (Marla *et al.* 2017). These findings suggest that a different closely linked gene, or the nearby *Tan1* gene, underlie this chilling tolerance QTL. No other QTL colocalized with orthologs of known plant cold tolerance genes (Thomashow 2001; Welti *et al.* 2002; Moellering *et al.* 2010).

The chilling tolerance QTL observed in our study may represent novel chilling tolerance mechanisms in sorghum, or conserved mechanisms not yet described in model plants. Fine-mapping and positional cloning of each chilling tolerance QTL (Ma *et al.* 2015) will be needed to address these or other hypotheses on the molecular basis of chilling tolerance in sorghum. Still, the genetic architecture provides some potential clues. Surprisingly, chilling tolerance QTL colocalized closely with classical tannin (*Tan1* and *Tan2*) and dwarfing genes (*Dw1* and *Dw3*) (Figure 3), four of the five most important genes under selection by US sorghum breeders in the 20th century (the fifth important gene, not colocalizing with chilling tolerance QTL is *Maturity1*) (Karper and Quinby 1946; Stephens *et al.* 1967; Wu *et al.* 2012; Morris *et al.* 2013a). This finding contradicted our original hypothesis of weak coupling-phase linkage of chilling susceptibility alleles with nontannin and dwarfing alleles. The colocalization itself could be due to (i) tight linkage (*e.g.*, <1 Mb) of chilling tolerance loci to classical tannin and dwarfing loci or (ii) pleiotropic effects of classical tannin and dwarfing loci on chilling tolerance.

First we considered whether coinheritance of tannin and chilling tolerance alleles could be due to a pleiotropic effect of seed pigmentation regulators (*Tan1* and *Tan2*) on chilling tolerance. Although BTx623 and Chinese parents all harbor functional *Tan2* alleles (based on complementary dominance of tannin genes), the chilling tolerance QTL near *Tan2* could be due to allelic variation at *Tan2* that does not alter grain tannins. A conserved MBW ternary complex controls

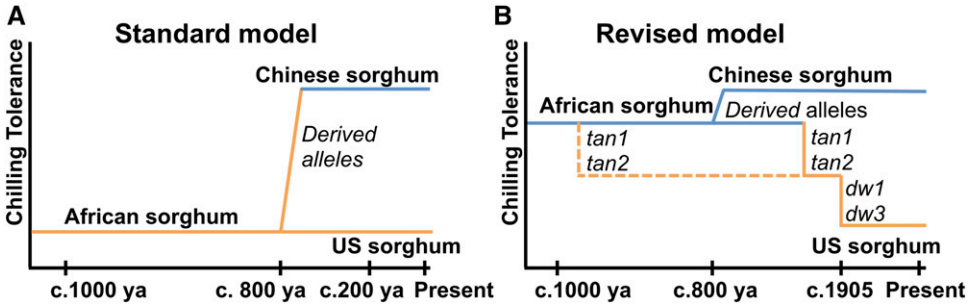

Figure 4 Evolutionary origin and agronomic effects of chilling tolerance (A) Standard model: African sorghums are chilling sensitive based on their tropical origin, sorghum dispersed into northern China (c. 800 years ago) has adapted to chilling while the US sorghums derived from African sorghums remain chilling-sensitive. (B) Based on the genetic architecture of early-season chilling tolerance, we revised the model to explain chilling sensitivity of US sorghums. Coinheritance of chilling tolerance loci with wildtype alleles of classical dwarfing (*Dw1* and *Dw3*) and tannin (*Tan1* and *Tan2*) genes suggest tropical-origin sorghums are chilling-tolerant. Inadvertent selection of chilling-sensitive alleles with favorable dwarfing (*dw1* and *dw3*) and nontannin (*tan1* and *tan2*) alleles resulted in persistence of chilling sensitivity in US sorghums, despite breeding for chilling tolerance over the past 50 years.

biosynthesis of flavonoids and tannins in plants via interactions of Myb and bHLH transcription factors with a WD40 transcriptional regulator (Nesi *et al.* 2000; Gu *et al.* 2011; Gao *et al.* 2018). Among sorghum tannin genes, *Tan1* encodes the WD40 component (Wu *et al.* 2012) and *Tan2* colocalizes with the bHLH transcription factor (Sobic.002G076600) (Morris *et al.* 2013b) orthologous to Arabidopsis *TRANSPARENT TESTA8* (*AtTT8*) and rice red grain gene (*OsRc*) (Nesi *et al.* 2000; Gu *et al.* 2011). The MBW complex has pleiotropic effects on abscisic acid-mediated seed dormancy and polyphenol-mediated protection from soil-borne pathogens (Helsper *et al.* 1994; Gu *et al.* 2011; Jia *et al.* 2012), which could contribute to emergence and seedling vigor under chilling. The chilling tolerance QTL *qSbCT02.08* detected in JLM of nontannin RILs (Figure S23) suggests that early-season chilling tolerance does not require seed tannins, even if the trait is under the control of the MBW complex. The existence of a Chinese accession Gai Gaoliang (PI 610727) that is chilling-tolerant but lacks grain tannins (Burow *et al.* 2010) supports this hypothesis.

Next we considered whether plant height alleles (*Dw1* and *Dw3*) could have pleiotropic effects on chilling tolerance that explain their colocalization with *qSbCT07.59* and *qSbCT09.57* (Figure S9). *Dw1*, which colocalized with *qSbCT09.57*, encodes a novel component of brassinosteroid (BR) signaling (Hirano *et al.* 2017). BR signaling controls cold tolerance mechanisms in tomato (Xia *et al.* 2018) and Arabidopsis (Eremina *et al.* 2016) so colocalization of *qSbCT09.57* with *Dw1* could reflect a pleiotropic chilling tolerance effect of DW1 BR signaling. *Dw3*, which colocalized with *qSbCT07.59*, encodes an auxin transporter. However, to our knowledge, no reports have demonstrated a role of auxin signaling in chilling tolerance.

Origins and consequences of the genetic architecture of chilling tolerance

Chilling sensitivity of US sorghum has generally been understood to be a result of sorghum's tropical origin (Stickler *et al.* 1962; Knoll *et al.* 2008) (Figure 4A), in keeping with a classic phytogeographic model (Vavilov 1951). Under this model, ancestrally chilling-sensitive African sorghums would have adapted to cold upon diffusion to temperate regions in central Asia and northern China (c. 800 years ago) due to derived alleles (Kimber 2000). However, our finding that chilling tolerance alleles coinherited with the ancestral wildtype alleles of classical tannin and dwarfing genes, which are widespread in both African and Chinese sorghums, suggests this model may be incorrect.

Instead, a revised model for derived chilling sensitivity of US sorghum and inadvertent selection may be more parsimonious (Figure 4B). Under this model, the African sorghums introduced into the US harbored basal chilling tolerance, but chilling sensitivity was inadvertently selected along with loss-of-function alleles at *tan1* and *tan2* (from African standing variation), and *dw1* and *dw3* (from *de novo* mutations in US) (Multani *et al.* 2003; Morris *et al.* 2013b; Hilley *et al.* 2016). Supporting this revised model, 38 RILs selected for agronomic suitability by the sorghum breeder (R.P.) were fixed for the chilling-susceptibility alleles (at *qSbCT09.57* and *qSbCT07.59*) that are coinherited with desired *dw1* and *dw3* alleles, respectively (Table S8). Thus, coinheritance of chilling susceptibility with desired traits likely stymied >50 years of chilling tolerance breeding in this crop (Stickler *et al.* 1962; Tiryaki and Andrews 2001; Yu and Tuinstra 2001; Knoll and Ejeta 2008; Burow *et al.* 2010; Kapanigowda *et al.* 2013).

A genotype-to-phenotype modeling approach, which couples genetic and ecophysiological modeling, can help assess the potential value of genotypes in a crop's target population of environments (Cooper *et al.* 2014). Preliminary ecophysiological modeling suggests that (were it not for chilling sensitivity) a standard grain sorghum hybrid could escape drought and have higher yields (~5%) if planted 30–60 days early (Figure S24). The improved power and resolution with the chilling NAM provides several new paths to obtain chilling tolerance while bypassing undesirable characteristics from Chinese sorghum. Several chilling tolerance alleles (at *qSbCT05.04*, *qSbCT07.10*, *qSbCT01.13*, and *qSbCT01.57*) are not coinherited with undesirable alleles for tannins and height (Figure 3) and can be used directly in marker-assisted introgression. Complementary dominance of *Tan1* and non-functional *tan2* (Wu *et al.* 2012) can be exploited to develop chilling-tolerant sorghums that retain the nontannin phenotype. If the standard model is correct (Figure 4A), rare recombinants identified with high-density markers will decouple chilling tolerance alleles from undesirable wildtype alleles of tannin and dwarfing genes and bypass undesirable coinheritance. If the revised model is correct (Figure 4B), antagonistic pleiotropic effects could be bypassed with novel tannin biosynthesis mutations to disrupt tannin production in *Tan1Tan2* chilling-tolerant background and novel dwarfing mutants (Jiao *et al.* 2016) in *Dw1Dw3* chilling-tolerant background.

Conclusions

Genetic tradeoffs caused by linkage drag have long been appreciated by geneticists and breeders (Zhu *et al.* 2018; Cobb *et al.* 2018). More recently, genetic tradeoffs due to antagonistic pleiotropy or conditional neutrality (Anderson *et al.* 2011) have been revealed by positional cloning of key agronomic genes (*i.e.*, those under strong selection in 20th century breeding programs). For instance, antagonistic pleiotropic

effects were identified for key improvement alleles of rice *semi-dwarf1* (Li *et al.* 2018) and tomato *jointless* (Soyk *et al.* 2017). In elite rice germplasm, conditional neutrality led to unintentional fixation of a drought-susceptibility allele at *Deeper rooting1* (Uga *et al.* 2013). Similarly, our findings suggest that strong selection for nontannin alleles (*tan1* and *tan2*) and dwarfing alleles (*dw1* and *dw3*) in grain sorghum in the 20th century inadvertently resulted in the loss of early-season chilling tolerance, due either (i) to tight repulsion-phase linkage of desired alleles (Figure 4A) or (ii) antagonistic pleiotropic effects of desired alleles on chilling susceptibility (Figure 4B). Given increasing evidence of genetic tradeoffs for genes under strong directional selection, characterizing both the genetic architecture and molecular basis of adaptive variation will be critical to guide genomics-enabled breeding and understand adaptive mechanisms.

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LITERATURE CITED

Anderson, J. T., J. H. Willis, and T. Mitchell-Olds, 2011 Evolutionary genetics of plant adaptation. Trends Genet. 27: 258–266. https://doi.org/10.1016/j.tig.2011.04.001

Araus, J. L., and J. E. Cairns, 2014 Field high-throughput phenotyping: the new crop breeding frontier. Trends Plant Sci. 19: 52–61. https://doi.org/10.1016/j.tplants.2013.09.008

Araya, A., I. Kisekka, P. H. Gowda, and P. V. V. Prasad, 2018 Grain sorghum production functions under different irrigation capacities. Agric. Water Manag. 203: 261–271.

Barrett, R. D. H., and D. Schluter, 2008 Adaptation from standing genetic variation. Trends Ecol. Evol. 23: 38–44. https://doi.org/10.1016/j.tree.2007.09.008

Bates, D., M. Mächler, B. Bolker, and S. Walker, 2015 Fitting linear mixed-effects models using lme4. J. Stat. Softw. 67. https://doi.org/10.18637/jss.v067.i01

Bernardo, R., 2008 Molecular markers and selection for complex traits in plants: Learning from the last 20 years. Crop Sci. 48: 1649. https://doi.org/10.2135/cropsci2008.03.0131

Bouchet, S., M. O. Olatoye, S. R. Marla, R. Perumal, T. Tesso *et al.*, 2017 Increased power to dissect adaptive traits in global sorghum diversity using a nested association mapping population. Genetics 206: 573–585. https://doi.org/10.1534/genetics.116.198499

Boyles, R. E., B. K. Pfeiffer, E. A. Cooper, B. L. Rauh, K. J. Zielinski *et al.*, 2017 Genetic dissection of sorghum grain quality traits using diverse and segregating populations. Theor. Appl. Genet. 130: 697–716. https://doi.org/10.1007/s00122-016-2844-6

Broman, K. W., H. Wu, S. Sen, and G. A. Churchill, 2003 R/qtl: QTL mapping in experimental crosses. Bioinformatics 19: 889–890. https://doi.org/10.1093/bioinformatics/btg112

Brown, P. J., W. L. Rooney, C. Franks, and S. Kresovich, 2008 Efficient mapping of plant height quantitative trait loci in a sorghum association population with introgressed dwarfing genes. Genetics 180: 629–637.

Browning, B. L., and S. R. Browning, 2013 Improving the accuracy and efficiency of identity-by-descent detection in population data. Genetics 194: 459–471. https://doi.org/10.1534/genetics.113.150029

Browning, B. L., and S. R. Browning, 2016 Genotype imputation with millions of reference samples. Am. J. Hum. Genet. 98: 116–126. https://doi.org/10.1016/j.ajhg.2015.11.020

Buckler, E. S., J. B. Holland, P. J. Bradbury, C. B. Acharya, P. J. Brown *et al.*, 2009 The genetic architecture of maize flowering time. Science 325: 714–718. https://doi.org/10.1126/science.1174276

Burow, G., J. J. Burke, Z. Xin, and C. D. Franks, 2010 Genetic dissection of early-season cold tolerance in sorghum (*Sorghum bicolor* (L.) Moench). Mol. Breed. 28: 391–402. https://doi.org/10.1007/s11032-010-9491-4

Cobb, J. N., P. S. Biswas, and J. D. Platten, 2018 Back to the future: revisiting MAS as a tool for modern plant breeding. Theor. Appl. Genet. https://doi.org/10.1007/s00122–018–3266–4

Cooper, M., C. D. Messina, D. Podlich, L. R. Totir, A. Baumgarten *et al.*, 2014 Predicting the future of plant breeding: complementing empirical evaluation with genetic prediction. Crop Pasture Sci. 65: 311–336. https://doi.org/10.1071/CP14007

Cramer, W., D. W. Kicklighter, A. Bondeau, B. M. Iii, G. Churkina *et al.*, 1999 Comparing global models of terrestrial net primary productivity (NPP): overview and key results. Glob. Change Biol. 5: 1–15. https://doi.org/10.1046/j.1365-2486.1999.00009.x

Danecek, P., A. Auton, G. Abecasis, C. A. Albers, E. Banks *et al.*, 2011 The variant call format and VCFtools. Bioinformatics 27: 2156–2158. https://doi.org/10.1093/bioinformatics/btr330

Dong, H., S. Yan, J. Liu, P. Liu, and J. Sun, 2019 TaCOLD1 defines a new regulator of plant height in bread wheat. Plant Biotechnol. J. 17: 687–699.

Elshire, R. J., J. C. Glaubitz, Q. Sun, J. A. Poland, K. Kawamoto *et al.*, 2011 A robust, simple genotyping-by-sequencing (gbs) approach for high diversity species. PLoS One 6: e19379. https://doi.org/10.1371/journal.pone.0019379

Eremina, M., S. J. Unterholzner, A. I. Rathnayake, M. Castellanos, M. Khan *et al.*, 2016 Brassinosteroids participate in the control of basal and acquired freezing tolerance of plants. Proc. Natl. Acad. Sci. USA 113: E5982–E5991. https://doi.org/10.1073/pnas.1611477113

Fiedler, K., W. A. Bekele, C. Matschegewski, R. Snowdon, S. Wieckhorst *et al.*, 2016 Cold tolerance during juvenile development in sorghum: a comparative analysis by genomewide association and linkage mapping. Plant Breed. 135: 598–606. https://doi.org/10.1111/pbr.12394

Franks, C. D., G. B. Burow, and J. J. Burke, 2006 A comparison of U.S. and Chinese sorghum germplasm for early season cold tolerance. Crop Sci. 46: 1371–1376. https://doi.org/10.2135/cropsci2005.08-0279

Gao, Y., J. Liu, Y. Chen, H. Tang, Y. Wang *et al.*, 2018 Tomato SlAN11 regulates flavonoid biosynthesis and seed dormancy by interaction with bHLH proteins but not with MYB proteins. Hortic. Res. 5: 27. https://doi.org/10.1038/s41438-018-0032-3

Glaubitz, J. C., T. M. Casstevens, F. Lu, J. Harriman, R. J. Elshire *et al.*, 2014 TASSEL-GBS: a high capacity genotyping by sequencing analysis pipeline. PLoS One 9: e90346. https://doi.org/10.1371/journal.pone.0090346

Grain sorghum production handbook, 1998 Grain Sorghum Prod. Handb. Kans. State Univ.

Gu, X.-Y., M. E. Foley, D. P. Horvath, J. V. Anderson, J. Feng *et al.*, 2011 Association between seed dormancy and pericarp color is controlled by a pleiotropic gene that regulates abscisic acid and flavonoid synthesis in weedy red rice. Genetics 189: 1515–1524. https://doi.org/10.1534/genetics.111.131169

Harlan, J. R., and J. M. J. de Wet, 1972 A simplified classification of cultivated sorghum. Crop Sci. 12: 172–176. https://doi.org/10.2135/cropsci1972.0011183X001200020005x

Helsper, J. P. F. G., A. V. Norel, K. Burger-Meyer, and J. M. Hoogendijk, 1994 Effect of the absence of condensed tannins in faba beans (*Vicia faba*) on resistance to foot rot, Ascochyta blight and chocolate spot. J. Agric. Sci. 123: 349–355. https://doi.org/10.1017/S0021859600070350

Hilley, J., S. Truong, S. Olson, D. Morishige, and J. Mullet, 2016 Identification of *Dw1*, a regulator of sorghum stem internode

length. PLoS One 11: e0151271. https://doi.org/10.1371/journal.pone.0151271

Hilley, J. L., B. D. Weers, S. K. Truong, R. F. McCormick, A. J. Mattison *et al.*, 2017 Sorghum *Dw2* encodes a protein kinase regulator of stem internode length. Sci. Rep. 7: 4616. https://doi.org/10.1038/s41598-017-04609-5

Hirano, K., M. Kawamura, S. Araki-Nakamura, H. Fujimoto, K. Ohmae-Shinohara *et al.*, 2017 Sorghum DW1 positively regulates brassinosteroid signaling by inhibiting the nuclear localization of BRASSINOSTEROID INSENSITIVE 2. Sci. Rep. 7: 126. https://doi.org/10.1038/s41598-017-00096-w

Holland, J. B., W. E. Nyquist, and C. T. Cervantes-Martínez, 2003 Estimating and interpreting heritability for plant breeding: an update, pp. 9–112 in *Plant Breeding Reviews*, edited by Janick, J. John Wiley & Sons, Inc., Hoboken, NJ.

Hu, Y., L. Jiang, F. Wang, and D. Yu, 2013 Jasmonate regulates the inducer of CBF expression-C-repeat binding factor/DRE binding factor1 cascade and freezing tolerance in Arabidopsis. Plant Cell 25: 2907–2924. https://doi.org/10.1105/tpc.113.112631

Hu, Z., M. O. Olatoye, S. Marla, and G. P. Morris, 2019 An integrated genotyping-by-sequencing polymorphism map for over 10,000 sorghum genotypes. Plant Genome 12: 0. https://doi.org/10.3835/plantgenome2018.06.0044

Jia, L., Q. Wu, N. Ye, R. Liu, L. Shi *et al.*, 2012 Proanthocyanidins inhibit seed germination by maintaining a high level of abscisic acid in Arabidopsis thaliana. J. Integr. Plant Biol. 54: 663–673. https://doi.org/10.1111/j.1744-7909.2012.01142.x

Jiao, Y., J. Burke, R. Chopra, G. Burow, J. Chen *et al.*, 2016 A sorghum mutant resource as an efficient platform for gene discovery in grasses. Plant Cell 28: 1551–1562. https://doi.org/10.1105/tpc.16.00373

Jones, J. W., G. Hoogenboom, C. H. Porter, K. J. Boote, W. D. Batchelor *et al.*, 2003 The DSSAT cropping system model. Eur. J. Agron. 18: 235–265.

Kapanigowda, M. H., R. Perumal, R. M. Aiken, T. J. Herald, S. R. Bean *et al.*, 2013 Analyses of sorghum [*Sorghum bicolor* (L.) Moench] lines and hybrids in response to early-season planting and cool conditions. Can. J. Plant Sci. 93: 773–784. https://doi.org/10.4141/cjps2012-311

Karper, R. E., and J. R. Quinby, 1946 The history and evolution of milo in the United States. Agron. J. 38: 441–453. https://doi.org/10.2134/agronj1946.00021962003800050007x

Kimber, C. T., 2000 Origins of domesticated sorghum and its early diffusion to India and China, pp. 3–98 in *Sorghum: origin, history, technology, and production*, edited by Smith, C. W., and R. A. Frederiksen. John Wiley and Sons, Hoboken, NJ.

Knight, M. R., and H. Knight, 2012 Low-temperature perception leading to gene expression and cold tolerance in higher plants. New Phytol. 195: 737–751.

Knoll, J., N. Gunaratna, and G. Ejeta, 2008 QTL analysis of early-season cold tolerance in sorghum. Theor. Appl. Genet. 116: 577–587. https://doi.org/10.1007/s00122-007-0692-0

Knoll, J., and G. Ejeta, 2008 Marker-assisted selection for early-season cold tolerance in sorghum: QTL validation across populations and environments. Theor. Appl. Genet. 116: 541–553. https://doi.org/10.1007/s00122-007-0689-8

Lasky, J. R., H. D. Upadhyaya, P. Ramu, S. Deshpande, C. T. Hash *et al.*, 2015 Genome-environment associations in sorghum landraces predict adaptive traits. Sci. Adv. 1: e1400218. https://doi.org/10.1126/sciadv.1400218

Li, W., M. Li, W. Zhang, R. Welti, and X. Wang, 2004 The plasma membrane-bound *phospholipase Ddelta* enhances freezing tolerance in *Arabidopsis thaliana*. Nat. Biotechnol. 22: 427–433. https://doi.org/10.1038/nbt949

Li, S., Y. Tian, K. Wu, Y. Ye, J. Yu *et al.*, 2018 Modulating plant growth–metabolism coordination for sustainable agriculture. Nature 560: 595–600. https://doi.org/10.1038/s41586-018-0415-5

Long, S. P., and A. K. Spence, 2013 Toward cool C4 crops. Annu. Rev. Plant Biol. 64: 701–722. https://doi.org/10.1146/annurev-arplant-050312-120033

Lyons, J. M., 1973 Chilling Injury in Plants. Annu. Rev. Plant Physiol. 24: 445–466. https://doi.org/10.1146/annurev.pp.24.060173.002305

Ma, Y., X. Dai, Y. Xu, W. Luo, X. Zheng *et al.*, 2015 COLD1 confers chilling tolerance in rice. Cell 160: 1209–1221. https://doi.org/10.1016/j.cell.2015.01.046

Mace, E. S., J.-F. Rami, S. Bouchet, P. E. Klein, R. R. Klein *et al.*, 2009 A consensus genetic map of sorghum that integrates multiple component maps and high-throughput Diversity Array Technology (DArT) markers. BMC Plant Biol. 9: 13. https://doi.org/10.1186/1471-2229-9-13

Maiti, R. K., P. S. Raju, and F. R. Bidinger, 1981 Evaluation of visual scoring for seedling vigour in sorghum. Seed Sci. Technol. Neth. 9: 613–622.

Mao, D., Y. Xin, Y. Tan, X. Hu, J. Bai *et al.*, 2019 Natural variation in the *HAN1* gene confers chilling tolerance in rice and allowed adaptation to a temperate climate. Proc. Natl. Acad. Sci. USA 116: 3494–3501. https://doi.org/10.1073/pnas.1819769116

McCormick, R. F., S. K. Truong, A. Sreedasyam, J. Jenkins, S. Shu *et al.*, 2018 The *Sorghum bicolor* reference genome: improved assembly, gene annotations, a transcriptome atlas, and signatures of genome organization. Plant J. https://doi.org/10.1111/tpj.13781

Menz, M. A., R. R. Klein, N. C. Unruh, W. L. Rooney, P. E. Klein *et al.*, 2004 Genetic diversity of public inbreds of sorghum determined by mapped AFLP and SSR markers. Crop Sci. 44: 1236. https://doi.org/10.2135/cropsci2004.1236

Meyer, R. S., and M. D. Purugganan, 2013 Evolution of crop species: genetics of domestication and diversification. Nat. Rev. Genet. 14: 840–852. https://doi.org/10.1038/nrg3605

Moellering, E. R., B. Muthan, and C. Benning, 2010 Freezing tolerance in plants requires lipid remodeling at the outer chloroplast membrane. Science 330: 226–228. https://doi.org/10.1126/science.1191803

Monk, R., C. Franks, and J. Dahlberg, 2014 Sorghum, pp. 293–310 in *Yield Gains in Major US Field Crops*, Crop Science Society of America.

Morris, G. P., P. Ramu, S. P. Deshpande, C. T. Hash, T. Shah *et al.*, 2013a Population genomic and genome-wide association studies of agroclimatic traits in sorghum. Proc. Natl. Acad. Sci. USA 110: 453–458. https://doi.org/10.1073/pnas.1215985110

Morris G. P., D. H. Rhodes, Z. Brenton, P. Ramu, V. M. Thayil, *et al.*, 2013b Dissecting genome-wide association signals for loss-of-function phenotypes in sorghum flavonoid pigmentation traits. G3 (Bethesda)3: 2085–2094. https://doi.org/10.1534/g3.113.008417

Multani, D. S., S. P. Briggs, M. A. Chamberlin, J. J. Blakeslee, A. S. Murphy *et al.*, 2003 Loss of an MDR transporter in compact stalks of maize *br2* and sorghum *dw3* mutants. Science 302: 81–84. https://doi.org/10.1126/science.1086072

Nesi, N., I. Debeaujon, C. Jond, G. Pelletier, M. Caboche *et al.*, 2000 The *TT8* gene encodes a basic helix-loop-helix domain protein required for expression of *DFR* and *BAN* genes in Arabidopsis siliques. Plant Cell 12: 1863–1878. https://doi.org/10.1105/tpc.12.10.1863

Nice, L. M., B. J. Steffenson, G. L. Brown-Guedira, E. D. Akhunov, C. Liu *et al.*, 2016 Development and genetic characterization of an advanced backcross-nested association mapping (AB-NAM) population of wild × cultivated barley. Genetics 203: 1453–1467. https://doi.org/10.1534/genetics.116.190736

Olsen, K. M., and J. F. Wendel, 2013 Crop plants as models for understanding plant adaptation and diversification. Front. Plant Evol. Dev. 4: 290. https://doi.org/10.3389/fpls.2013.00290

Orr, H. A., 2005 The genetic theory of adaptation: a brief history. Nat. Rev. Genet. 6: 119–127. https://doi.org/10.1038/nrg1523

Ortiz, D., J. Hu, and M. G. Salas Fernandez, 2017 Genetic architecture of photosynthesis in *Sorghum bicolor* under non-stress and cold stress conditions. J. Exp. Bot. 68: 4545–4557. https://doi.org/10.1093/jxb/erx276

Paaby, A. B., and M. V. Rockman, 2013 The many faces of pleiotropy. Trends Genet. 29: 66–73. https://doi.org/10.1016/j.tig.2012.10.010

Paradis, E., J. Claude, and K. Strimmer, 2004 APE: analyses of phylogenetics and evolution in R language. Bioinformatics 20: 289–290.

Park, S., C.-M. Lee, C. J. Doherty, S. J. Gilmour, Y. Kim *et al.*, 2015 Regulation of the Arabidopsis CBF regulon by a complex

low-temperature regulatory network. Plant J. 82: 193–207. https://doi.org/10.1111/tpj.12796

Paterson, A. H., J. E. Bowers, R. Bruggmann, I. Dubchak, J. Grimwood *et al.*, 2009 The *Sorghum bicolor* genome and the diversification of grasses. Nature 457: 551–556. https://doi.org/10.1038/nature07723

Quinby, J. R., and R. E. Karper, 1954 Inheritance of height in sorghum. Agron. J. 46: 211–216.

R Core Team, 2018 *R: A Language and Environment for Statistical Computing*. Vienna, Austria.

Soyk, S., Z. H. Lemmon, M. Oved, J. Fisher, K. L. Liberatore *et al.*, 2017 Bypassing negative epistasis on yield in tomato imposed by a domestication gene. Cell 169: 1142–1155.e12. https://doi.org/10.1016/j.cell.2017.04.032

Staggenborg, S. A., and R. L. Vanderlip, 2005 Crop simulation models can be used as dryland cropping systems research tools. Agron. J. 97: 378–384.

Stephens, J. C., 1946 A second factor for subcoat in sorghum seed. Agron. J. 38: 340–342.

Stephens, J. C., F. R. Miller, and D. T. Rosenow, 1967 Conversion of alien sorghums to early combine genotypes. Crop Sci. 7: 396. https://doi.org/10.2135/cropsci1967.0011183X000700040036x

Stickler, F. C., A. W. Pauli, and A. J. Casady, 1962 Comparative responses of kaoliang and other grain sorghum types to temperature. Crop Sci. 2: 136. https://doi.org/10.2135/cropsci1962.0011183X000200020015x

Thomashow, M. F., 2001 So What's new in the field of plant cold acclimation? lots! Plant Physiol. 125: 89–93. https://doi.org/10.1104/pp.125.1.89

Tiryaki, I., and D. J. Andrews, 2001 Germination and seedling cold tolerance in sorghum: II. Parental lines and hybrids. Agron. J. 93: 1391. https://doi.org/10.2134/agronj2001.1391

Tuberosa, R., 2012 Phenotyping for drought tolerance of crops in the genomics era. Front. Physiol. 3. https://doi.org/10.3389/fphys.2012.00347

Uga, Y., K. Sugimoto, S. Ogawa, J. Rane, M. Ishitani *et al.*, 2013 Control of root system architecture by *DEEPER ROOTING 1* increases rice yield under drought conditions. Nat. Genet. 45: 1097–1102. https://doi.org/10.1038/ng.2725

Vavilov N. I., 1951 The origin, variation, immunity and breeding of cultivated plants, translated from the Russian by K. Starr Chester. Chron. Bot.

Wang, X., L. Wang, Y. Wang, H. Liu, D. Hu *et al.*, 2018 Arabidopsis *PCaP2* plays an important role in chilling tolerance and ABA response by activating CBF- and SnRK2-mediated transcriptional regulatory network. Front. Plant Sci. 9. https://doi.org/10.3389/fpls.2018.00215

Welti, R., W. Li, M. Li, Y. Sang, H. Biesiada *et al.*, 2002 Profiling membrane lipids in plant stress responses role of *PHOSPHOLIPASE Dα* in freezing-induced lipid changes in Arabidopsis. J. Biol. Chem. 277: 31994–32002. https://doi.org/10.1074/jbc.M205375200

White, J. W., G. Alagarswamy, M. J. Ottman, C. H. Porter, U. Singh *et al.*, 2015 An overview of CERES–sorghum as implemented in the cropping system model version 4.5. Agron. J. 107: 1987–2002.

Wu, Y., X. Li, W. Xiang, C. Zhu, Z. Lin *et al.*, 2012 Presence of tannins in sorghum grains is conditioned by different natural alleles of *Tannin1*. Proc. Natl. Acad. Sci. USA 109: 10281–10286. https://doi.org/10.1073/pnas.1201700109

Xia, X.-J., P.-P. Fang, X. Guo, X.-J. Qian, J. Zhou *et al.*, 2018 Brassinosteroid-mediated apoplastic H2O2-glutaredoxin 12/14 cascade regulates antioxidant capacity in response to chilling in tomato. Plant Cell Environ. 41: 1052–1064. https://doi.org/10.1111/pce.13052

Xiang, W., 2009 Identification of two interacting quantitative trait loci controlling for condensed tannin in sorghum grain and grain quality analysis of a sorghum diverse collection [M. Sc.]: Kansas State University.

Yu, J., and M. R. Tuinstra, 2001 Genetic analysis of seedling growth under cold temperature stress in grain sorghum. Crop Sci. 41: 1438–1443. https://doi.org/10.2135/cropsci2001.4151438x

Zeng, Z. B., 1994 Precision mapping of quantitative trait loci. Genetics 136: 1457–1468.

Zhang, C., S.-S. Dong, J.-Y. Xu, W.-M. He, and T.-L. Yang, 2018 PopLDdecay: a fast and effective tool for linkage disequilibrium decay analysis based on variant call format files. Bioinformatics. https://doi.org/10.1093/bioinformatics/bty875

Zhu, G., S. Wang, Z. Huang, S. Zhang, Q. Liao *et al.*, 2018 Rewiring of the fruit metabolome in tomato breeding. Cell 172: 249–261.e12. https://doi.org/10.1016/j.cell.2017.12.019

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