Integrated metabolomics and transcriptomic analysis of the flavonoid regulatory networks in *Sorghum bicolor* seeds

Yaxing Zhou1, Jingbo Lv2, Zhonghao Yu1, Zhenguang Wang2, Yan Li2, Mo Li2, Zhilan Deng2, Qingquan Xu2, Fengjuan Cui2 and Wei Zhou1*

**Abstract**

**Background:** The objective of this study was to reveal the flavonoid biosynthesis pathway in white (Z6), red (Z27) and black (HC4) seeds of the sweet sorghum (*Sorghum bicolor*) using metabolomics and transcriptomics, to identify different flavonoid metabolites, and to analyze the differentially expressed genes involved in flavonoid biosynthesis.

**Results:** We analyzed the metabolomics and transcriptomics data of sweet sorghum seeds. Six hundred and fifty-one metabolites including 171 flavonoids were identified in three samples. Integrated analysis of transcriptomics and metabolomics showed that 8 chalcone synthase genes (*gene19114, gene19115, gene19116, gene19117, gene19118, gene19120, gene19122 and gene19123*) involved in flavonoid biosynthesis, were identified and play central role in change of color. Six flavanone including homoeriodictyol, naringin, prunin, naringenin, hesperetin and pinocembrin were main reason for the color difference.

**Conclusions:** Our results provide valuable information on the flavonoid metabolites and the candidate genes involved in the flavonoid biosynthesis pathway in sweet sorghum seeds.

**Keywords:** Flavonoid, Transcriptome, Metabolite, Sweet sorghum, Seed

**Introduction**

Sweet sorghum (*Sorghum bicolor*) is one of the most important cereal crops in the world and is remarkable for its resistant to stress and extremely high photosynthetic efficiency. Therefore, sweet sorghum could achieve high yields when it widely grown in harsh environments such as arid or semi-arid areas [1]. As a C4 drought resistant species, sweet sorghum has a huge potential for bioenergy [2], and is considered to be an ideal candidate for biofuel production due to its high efficiency of photosynthesis and the ability to maintain yield under harsh environmental conditions [3]. The harvest sweet sorghums used for industrial ethanol and densified biofuel production could combine soil remediation with creating economic benefit [4].

In our research, the seeds of three cultivars are selected. These three colors are the representative colors of all sorghum colors, that is, other sorghum colors are transition colors of these three. We know that the comprehensive agronomic traits of these three colors of sorghum are better. Previous studies found that the content and type of anthocyanin are major determinant of seed color, which is important natural colorants, and is widely found in leaves, flowers, fruits, and roots [5, 6]. The anthocyanins are beneficial to plants from UV radiation and pathogen, and are a natural antioxidants...
Furthermore, anthocyanin play an important role in antioxidant properties and enhanced nutritional value for plant seed [8]. Meanwhile, anthocyanins contributes to the seed pigmentation, and the high level of anthocyanin on top of phenolic content contributes to the greater pharmacological capacities [9]. Model species have been used to identify functional and regulatory genes involved in the flavonoid biosynthesis pathway in which two main classes of functional enzymes were identified. The “early biosynthetic step” enzymes can catalyze all flavonoid synthesis, and “later biosynthetic step” enzymes play important role in synthesis of flavonols and anthocyanins [10, 11].

In the current study, our aim was to identify significantly differentially expressed genes and metabolites involved in flavonoid biosynthesis pathway in the seeds of three sweet sorghum cultivars. Furthermore, we conducted research on the regulatory networks of flavonoid biosynthesis in seeds of sweet sorghum using metabolomics and transcriptomics. Our results not only provide candidate genes but also valuable information for metabolic engineering of flavonoid biosynthesis in seeds of sweet sorghum.

Materials and methods

Plant materials and sample preparation

The source of sweet sorghum cultivars (namely Z6, Z27, and HC4) from Agricultural College of Inner Mongolia Minzu University (N42°15'-45°41', E119°15'-123°43') in Tongliao City (Inner Mongolia, China) with excellent quality characteristics were chosen. Seeds of these three cultivars are white (Z6), red (Z27) and black (HC4), respectively. The experiments were conducted at the Experimental Base of the Agricultural College of Inner Mongolia Minzu University (N42°15'-45°41', E119°15'-123°43'). The trial used a randomized block design with 3 replicates in planting 16 rows per plot. The sweet sorghum is bagged at the flowering stage, and mesh bag is eluted within this period [14, 15].

Seeds for each cultivar were collected from 5–6 plantlets as one sample. Each sample was split into two equal parts: one-half was used for metabolomic analysis and the other half for transcriptomic analysis. All the materials were frozen in liquid nitrogen immediately and then stored at -80 °C until use.

Metabolite extraction

The frozen seeds were crushed using a bead beater (1.5 min, 30 Hz, three repetitions, MM 400, Retsch). One hundred milligrams of the powdered sample was extracted overnight at 4 °C using 1 mL of 70% aqueous methanol containing 0.1 mg.L−1 lidocaine. After centrifugation at 10 000 g for 10 min, the supernatants were filtrated by using 0.22 μm hydrophilic poly(tetrafluoroethylene) syringe filters (SCAA-104, ANPEL, Shanghai, China) (http://www.anpel.com.cn/) before metabolomics analysis [13]. The quality control samples (mix1-3) were injected every three experimental samples throughout the analytical run to provide a set of data from which repeatability could be assessed.

Metabolite profiling using liquid Chromatography-Electrospray Ionization-Tandem Mass Spectrometry (LC–ESI–MS/MS)

Metabolite profiling was conducted using a LC–ESI–MS/MS system (HPLC, UFLC SHIMADZU CBM30A system; MS, Applied Biosystems 4500 Q TRAP) (http://www.shimadzu.com.cn/) and an Agilent 6520 accurate-mass time-of-flight mass spectrometer (http://www.appliedbiosystems.com.cn/). Chromatographic separation was performed on an ACQUITY UPLC HSS T3 C18 column (2.1 mm × 100 mm × 1.8 μm; Waters) using mobile phase A (0.04% acetic acid in deionized water) and mobile phase B (0.04% acetic acid in acetonitrile). The elution profile was used as follows: 95:5 v(A)/v(B) at 0 min, 5:95 v(A)/v(B) at 11.0 min, 5:95 v(A)/v(B) at 12.0 min, 95:5 v(A)/v(B) at 12.1 min, and 95:5 v(A)/v(B) at 15.0 min. The flow rate was maintained at 0.4 mL/min. Mass data acquisition was performed in electrospray ionization positive/negative mode using the following parameters: ion spray voltage of (±) 5.5 kV; ion source gas I of 55 psi; gas II of 60 psi; curtain gas of 25 psi; turbo spray temperature of 550 °C. Instrument tuning and mass calibration were performed with 10 and 100 μmol/L polypropylene glycol solutions in triple quadrupole and linear ion trap modes, respectively. Declustering potential (DP) and collision energy (CE) for individual multiple reaction monitoring (MRM) transitions were performed with specific DP and CE optimization. A specific set of MRM transitions were monitored for each period based on the metabolites eluted within this period [14, 15].

Qualitative and quantitative analysis of metabolites

To facilitate the identification of metabolites by widely targeted metabolomics approach (MetWare, Wuhan, China), accurate m/z value of each precursor ions (Q1) were obtained [16]. This method has been previously described [14]. In brief, metabolites were identified by comparing the m/z values, the retention time (RT), and the fragmentation patterns with the standards in a self-compiled database (MetWare). Significantly changed metabolites (SCMs) were filtered according to |Log2 (fold change)| ≥ 1, p-value < 0.05.
RNA sequencing

RNA isolation, purification and monitoring, cDNA library construction and sequencing were performed as previously described [17]. Briefly, RNA purity, concentration and integrity were checked, measured and assessed using the NanoPhotometer® spectrophotometer (IMPLEN, Westlake Village, CA, USA), Qubit® RNA Assay Kit in Qubit® 2.0 Flurometer (Life Technologies, Carlsbad, CA, USA) and RNA Nano 6000 Assay Kit of the Agilent Bioanalyzer 2100 system (Agilent Technologies, Santa Clara, CA, USA), respectively [18]. Sequencing libraries were generated using NEBNext® Ultra™ RNA Library Prep Kit for Illumina® (NEB, Ipswich, MA, USA) following manufacturer’s recommendations and were sequenced on an Illumina Hiseq platform. According to the manufacturer’s instructions, NEBNext® Ultra™ RNA Library Prep Kit for Illumina® (NEB, Ipswich, MA, USA) was used to generate sequencing libraries, which were then sequenced on an Illumina Hiseq platform [18]. The clean data of RNA-seq were available from National Center for Biotechnology Information Sequence Read Archive database (SRA accession numbers: SRR8662425, SRR8662424).

RT-qPCR (Real-Time Quantitative PCR)

The testing of RNA quality and determination of RNA concentration were performed by 1.0% agarose gel electrophoresis and micro ultraviolet spectrophotometry (Thermo NanoDrop 2000, Thermo Fisher Scientific, Waltham, MA, USA), respectively. Approximately 1 µg of total RNA was determined for cDNA synthesis using RevertAid™ First Strand cDNA synthesis kit (Thermo Fisher Scientific, Waltham, MA, USA). The LightCycler® 480 real-time PCR system with a 96-well plate was used to conduct an amplified reaction consisting of 95 °C for 5 min, followed by 45 cycles of 10 s at 95 °C, 20 s at 60 °C, and 20 s at 72 °C in a volume of 10 µL. At the end of each experiment, a melt-curve analysis was carried out using the default parameters (5 s at 95 °C and 1 min at 65 °C). The β-actin was used for normalization [19, 20]. All analyses were repeated three times using biological replicates.

Integrative analysis of metabolome and transcriptome

Metabolites and DEGs involved in phenylpropanoids biosynthesis and lipids metabolism in KEGG pathways were selected for integrative analysis. Metabolites used for correlation analysis were filtered according to variable importance in the project (VIP) > 1, p-value < 0.05, and |Log2 (Fold Change)| ≥ 1. Pearson correlation coefficients and p-values were calculated for metabolome and transcriptome data integration using the spearman method [21].

Statistical analysis

The relative expressions were calculated using the 2−\(\Delta\Delta C_t\) method [22], and GraphPad Prism 5 (GraphPad Software Inc., San Diego, CA, USA) was used for chart preparation. The R (www.rproject.org/) and MEGA6 were used to conduct the heatmap and cluster analysis. Principle component analysis (PCA) was performed by R to study gene variety-specific accumulation. IBM SPSS Statistics 20 was used to test significant differences.

Results

Morphological differences among three sweet sorghum cultivars

Though the three sweet sorghum cultivars were planted simultaneously and grown in the same field and under the same conditions, there were obvious differences in the morphology of three cultivars, especially the color of the seeds.

Z6 was white, and Z27 and HC4 possess red/black (Fig. 1). Meanwhile, the research indicated that spike length, shank length and cutin rate in HC4 were higher than Z6 and Z27.

Fig. 1  Seed color of the three sweet sorghum cultivars
Comprehensive analysis of metabolites
In order to explore the differences of metabolites in three sweet sorghum cultivars seeds (Fig. 1), a metabo-
lome program was carried out. Six hundred and fifty-one metabolites were identified in three samples, includ-
ing 34 flavonol, 58 flavone, 4 proanthocyanidins, 64 organic acids, 13 anthocyanins, and 29 amino acids. Then, according to the quantitative results of the identi-
fied metabolites, the differential metabolites between different group comparisons were analyzed based on fold-change and $p$-value. In the comparison between Z27 and HC4, a total of 111 and 106 metabolites presented as being up-regulated and down-regulated. Compared with HC4, Z6 had 182 metabolites up-regulated and 58 metabolites down-regulated. Compared with Z27, Z6 had 135 metabolites up-regulated and 54 metabolites down-regulated, respectively (Fig. 2A).

Analysis of anthocyanins
Flavonoids, including anthocyanins, flavanes, flavones, flavanones, flavonols are found in numerous colored fruits and play an important role in pigmentation [23]. One hundred seventy-one flavonoids were found in three samples and many of flavonoids were differen-
tially expressed in each comparison group. Seventy-five flavonoids were differentially expressed between Z6 vs. HC4. In HC4, procyanidin A1 and Procyanidin A2 were found with 24,740- and 5429- fold increments compared to Z6. Meanwhile, compared with Z6, 8 anthocyanins including, cyanidin 3,5-o-diglucoside (cyanin), cyanidin 3-o-glucosyl-malonylglucoside, cyanidin o-malonyl-malonylhexoside, cyanidin o-syringic acid, cyanidin o-acetylhexoside, cyanidin o-diacetyl-hexoside-o-glyceric acid, cyanidin 3-o-glucoside (kuromanin) and pelargonidin 3-o-beta-D-glucoside (callistephin chloride), significantly highly expressed in HC4.

Similarly, compared with Z6, procyanidin A3, procyanidi-
in A2, procyanidin B2 and procyanidin B3 were increased 820,740-, 28,851-, 1958-, and 1,466,666- fold in the Z27 samples. Furthermore, 6 anthocyanins were identified and differen-
tially expressed, of which peonidin O-hexoside, cyanidin o-syringic acid, delphinidin, cyanidin 3,5-o-diglu-
coside (cyanin) and pelargonidin 3-o-beta-d-glucoside (callistephin chloride) were up-regulated, and only delphinidin 3-o-glucoside (mirtilin) was down-regulated.

Ten anthocyanins were identified and differentially expressed in Z27 vs. HC4. Compared with Z27, peoni-
din, delphinidin and cyanidin 3,5-O-diglucoside (cy-
nin) were evidently down-regulated in HC4 group. Seven other anthocyanins were up-regulated in HC4.

Flavonol, flavone, and flavanone
Fourteen flavones, 15 flavonols and 9 flavanones were differen-
tially expressed in Z6 vs. Z27, among which syringetin 7-o-hexoside, dihydromyricetin, luteolin o-eudesmic acid-o-hexoside, myricetin showed 188,518-, 23,444-, 9088- and 8433-fold increases in Z27 with maximal difference. Twenty-one flavones, 12 flavonols and 10 flavanones were screened and differentially expressed in Z6 vs. HC4. Biocha-
nin, myricetin and quercetin 5-o-malonylhexosyl-hexoside

Fig. 2 Venn Diagram result in three pairs of comparison groups. Differentially accumulated metabolites in three pairs of comparison groups
contents were 45,111-, 30,962- and 26,740-fold higher, respectively, in the HC4 vs. Z6.

In Z27 and HC4 comparison group, we identified 19 flavonol in which only myricetin, kumatakenin Rhamnetin (7-O-methyl quercetin), and fustin were up-regulated in HC4 group. Twenty-nine flavone were found in HC4, most of which were up-regulated in HC4. Six flavanone were screened in HC4, in which eriodictyol and hesperetin 5-O-glucoside were up-regulated and others 4 flavanone were down-regulated.

Furthermore, we screened 16 DEMs from 45 common DEMs, including 1 procyanidine, 5 flavonol, 2 isoflavone, 3 flavanone, 4 flavone, and 1 anthocyanin, which were differentially expressed in three groups. The DEMs maybe play important roles in difference of color (Fig. 2B).

**Metabolic pathways:** Flavonoid biosynthesis and flavone and flavonol biosynthesis

Principal component analysis (PCA) was performed indicating that the score plots of PCA (Fig. 3) exhibited an

![PCA scores plot of different cultivars. PC1, principal component 1; PC2, principal component 2. Explained variants PC1: 62.7%, PC2: 25.1%](image)

![Numbers of differentially expressed genes](image)
obvious separation in 3 cultivars. The two principal components (PC1 and PC2) were 62.7% and 25.1%, respectively. The PCA clusters indicated that the three cultivars were distinct. Then, we compared the differentially expressed genes (DEGs) in every two cultivars. The number of differentially expressed genes had very high variance among different cultivars. In total, 3054 DEGs were identified including 1458 up-regulated and 1596 down-regulated genes, which were differentially expressed at Z6 vs. HC4. There were 1375 up-regulated DEGs and 1312 down-regulated DEGs at Z27 vs. HC4, 422 up-regulated DEGs and 450 down-regulated DEGs in Z6 vs. Z2 (Fig. 4).

**Table 1** KEGG enrichment pathways in different comparison groups

| Comparison group | Pathway                             | ko_ID  | DEG | gene ID                        | up/down |
|------------------|-------------------------------------|--------|-----|--------------------------------|---------|
| Z6_vs_Z27       | Flavone and flavonol biosynthesis  | ko00944| 1   | gene16076                      | Down    |
| Z6_vs_Z27       | Flavonoid biosynthesis              | ko00941| 3   | gene16076;gene24313;gene344    | Down    |
| Z27_vs_HC4      | Flavone and flavonol biosynthesis  | ko00944| 4   | Sorghum_bicolor_newGene_10649;gene16073;gene16076;gene16077;gene17170;gene19114;gene19115;gene19116;gene19117;gene19118;gene19120;gene19122;gene19123;gene26712;gene30278;gene344;gene6640;gene7823 | Up      |
| Z27_vs_HC4      | Flavonoid biosynthesis              | ko00941| 21  | Sorghum_bicolor_newGene_10137;Sorghum_bicolor_newGene_10649;gene14509;gene15470;gene16073;gene16076;gene16077;gene17170;gene19114;gene19115;gene19116;gene19117;gene19118;gene19120;gene19122;gene19123;gene26712;gene30278;gene344;gene6640;gene7823 | Up      |
| Z6_vs_HC4       | Flavone and flavonol biosynthesis  | ko00944| 1   | Sorghum_bicolor_newGene_10649;Sorghum_bicolor_newGene_10137;Sorghum_bicolor_newGene_10649;gene15470;gene17170;gene19114;gene19115;gene19116;gene19117;gene19118;gene19120;gene19122;gene19123;gene26712;gene30278;gene344 | Up      |
| Z6_vs_HC4       | Flavonoid biosynthesis              | ko00941| 14  | Sorghum_bicolor_newGene_10137;Sorghum_bicolor_newGene_10649;gene15470;gene17170;gene19114;gene19115;gene19116;gene19117;gene19118;gene19120;gene19122;gene19123;gene26712;gene30278;gene344 | Up      |

**Fig. 5** Gene and metabolite association analysis. In the ko00941 pathway, Gene and metabolite association analysis in Z6 vs. HC4 (A), Z6 vs.Z27 (B), and Z27 vs. HC4 (C). In ko00944 pathway, Gene and metabolite association analysis in Z6 vs. HC4 (D), Z6 vs. Z27 (E) and Z6 vs.HC4 (F).
KEGG analysis revealed flavonoid biosynthesis and flavone and flavonol biosynthesis as the significantly changed pathways. Furthermore, the related DEGs in the pathway were screened in Z6 vs. Z27, Z27 vs. HC4 and Z6 vs. HC4 (Table 1).

**Correlation network graph analysis genes and metabolites**

Subsequently, the DEGs and DEMs in flavonoid biosynthesis and flavone and flavonol biosynthesis were screened. Pearson correlation coefficient analysis was performed on the genes and metabolites to explore regulatory mechanisms. In the ko00941 pathway, 15 genes were related to 11 metabolites between Z6 and HC4 (Fig. 5A). Accordingly, gene 24313, gene16076 and gene344 were screened, which was relevant to 15 metabolites between Z6 and HC27 (Fig. 5B). 14 metabolites were related to 18 genes in Z27 vs. HC4 (Fig. 5C).

In ko00944 pathway, Sorghum_bicolor_newGene_10649 was found, and it was relevant to 5 metabolites including pme1541, pme6389, pme1478, pme2954 and pma1116 in Z6 vs. HC4 (Fig. 5D). The gene16076 was negative correlation with 5 metabolites in Z6 vs. Z27 (Fig. 5E). Fifteen genes were related to 11 metabolites between Z6 and HC4 (Fig. 5F).

**Regulatory network of flavonoid biosynthesis**

The color difference was greatest between Z6 and HC4, so the pathway of flavonoid biosynthesis were analyzed in two groups (Fig. 6A). Eight chalcone synthase
genes (gene19114, gene19115, gene19116, gene19117, gene19118, gene19120, gene19122 and gene19123) involved in flavonoid biosynthesis, were identified and up-regulated in HC4 (Fig. 6B). Caffeoyl-CoA O-methyltransferase 1 and flavanone 3-hydroxylase showed two-fold and 2.7-fold upregulation in the HC4 group. Six flavanone including homoeiodictyol, naringin, prunin, naringenin, hesperetin and pinocembrin were high fold up-regulation in HC4 (Fig. 6C). Myricetin was one of the most significantly DEMs in the HC4 vs. Z6 group, increasing 30,962-fold. 1 flavone (butin) was upregulated 1.4-fold in HC4 vs. Z6.

**RT-qPCR validation of the transcriptomic data**

To validate the key RNA-Seq results, we selected 8 genes and analyzed their expression levels in Z6, HC4 and Z27 using RT-qPCR. The results validated the relevance of the RNA-Seq data and RT-qPCR showed good consistency for both up- and down-regulated gene expression (Fig. 7).

**Discussion**

The color of the sweet sorghum seed of different cultivars varies widely, mainly due to the fact that the seed coat contains different flavonoids. A combined metabolome and transcriptome study can provide us with new, large-scale information on the shifted flavonoids and the underlying modifications in gene-expression networks.

In our research, integrated metabolomics and transcriptomic analysis were employed to explore gene expression and flavonoid alterations for the color of the sweet sorghum seed of different cultivars. The results indicated that there were significant differences in the seed color of three cultivars, grown in the same conditions. Meanwhile, 651 metabolic components were identified including 34 flavonol, 58 flavone, 4 proanthocyanidins and 13 anthocyanins, which are commonly identified in pink, red, purple and other deep-colored fruit [24]. To understand the metabolic characteristics of the color of the sweet sorghum seed, our attention was focused on main flavonoid compounds in all samples.

Our results indicated that 171 flavonoids were found in three samples. 75 flavonoids were differentially expressed between Z6 vs. HC4. The procyanidin A1 and procyanidin A2 were evidently higher in HC4 compared to Z6. Eight anthocyanins including were significantly highly expressed in HC4. Meanwhile, the contents of procyanidin A3, procyanidin A2, procyanidin B2 and procyanidin B3 were higher in the Z27 samples. Flavonoids are composed of anthocyanins, flavonoids, chalcone, flavonoids, flavonols and iso-flavones. Among them, anthocyanins are the main determinant of color formation in flower and fruit [25]. Previous study found that the cyanidin-3-O-rhamnoglucoside (cyanidin-3-O-rutinoside) was the main anthocyanin in the peel of “Black Mission,” “Bursa” and “Brown Turkey” figs [26, 27]. Acyl-modified anthocyanins have been identified in Arabidopsis [28]. Study indicated that the cyanidin 3-O-(malonyl)-glucoside was the only pigment responding to temperature in the cool-cultivated red lettuce [29]. Anthocyanins and flavonoids affect fruit color and taste [30]. We speculated that the darker color of seeds in Z27 and HC4 were mainly due to the higher anthocyanins content of these two groups.

![Fig. 7](image-url) The expression of representative genes in 3 samples validated by qRT-PCR
In the present study, several critical genes were highly expressed in HC4, which play important roles in the pathway of flavonoid biosynthesis. Eight chalcone synthase genes (CHS) involved in flavonoid biosynthesis were identified and up-regulated in HC4. Chalcone synthase, a key enzyme in the flavonoid biosynthetic pathway, has been studied in many plants [31]. The structural gene expression levels of chalconesynthase (CHS) and chalcone isomerase (CHI) were highly similar and significantly positively correlated with anthocyanin accumulation rate in wild Lycium ruthenicum Murr [32]. Consistent with our study, in crabapple cultivars with dark red, pink and white petal colors, CHS play an important role in the formation of the red coloration [33]. In the white-flowered individuals, the expression of CHS gene was significantly inhibited, while the expression of other genes in the anthocyanin biosynthesis pathway was similar to that in pigmented individuals [34].

In summary, our study provides insight on modulated anthocyanin and flavonoid expression in three sweet sorghum cultivars, revealing the large-scale changes in nutritionally important compounds and gene expression in seeds of three cultivars. Our results provide new information on the anthocyanidin, flavonol and procyanidin metabolites of sweet sorghum cultivars and the global transcriptional changes in seeds color regulation.

Acknowledgements
The authors sincerely thank the Natural Science Foundation of Inner Mongolia Autonomous Region for providing the facilities. The authors are grateful to the reviewers for their helpful comments and suggestions.

Guidelines statement
Our research complied with relevant institutional, national, and international guidelines and legislation.

Authors’ contributions
Wei Zhou as the corresponding author contributed to the conception of the study and contributed significantly to analysis; Yaxing Zhou as the first author wrote the manuscript; Jingbo Lv as the second author performed the experiments and manuscript preparation; Zhonghao Yu as the third author performed the data analyses; Zhenruo Wang and Yan Li as the fourth and fifth authors helped perform the analysis with constructive discussions. Mo Li, Zhihan Deng, Qingquan Xu and Fengjuan Cui as the remaining co-authors contributed to genomic and transcriptomic data analysis and interpretation, figure design, and edited the manuscript. All authors read and approved the final manuscript.

Funding
This research was supported by Inner Mongolia Science and Technology Innovation Guide Project (KJCX1702), Project of the Education Department of Inner Mongolia Autonomous Region (N2017193), Research Project of Inner Mongolia Minzu University (BS621); Natural Science Foundation of Inner Mongolia Autonomous Region (2019MS03076), Inner Mongolia Autonomous Region Science and Technology Plan (2020GG0102), Inner Mongolia Autonomous Region College Youth Science and Technology Talent Project (NJYT20252), Research Project on Germplasm Identification of Salt-Alkali-tolerant High-quality Sorghum in West Liaobe Plain, National Ethnic Affairs Commission Higher Education Teaching Reform Research Project (21087) and Research Project of Inner Mongolia Minzu University (NMDYB18036).

Availability of data and materials
The datasets generated and/or analyzed during the current study are available in National Center for Biotechnology Information Sequence Read Archive database (https://www.ncbi.nlm.nih.gov/sra/?term, SRA accession numbers: SRR8662425, SRR8662424).

Declarations

Ethics approval and consent to participate
Our research complies with relevant institutional, national, and international guidelines and legislation.

Consent for publication
Not applicable.

Competing interests
The authors declare no conflicts of interest.

Author details
1 Agricultural College, Inner Mongolia Minzu University, No. 996 Xilamulun Street, Keriqin District, Tongliao 028000, Inner Mongolia, People’s Republic of China. 2 Tongliao Agriculture and Animal Husbandry Research Institute, Tongliao 028000, Inner Mongolia, People’s Republic of China.

Received: 7 March 2022 Accepted: 17 August 2022
Published online: 26 August 2022

References
1. Zhao Y, Lu Z, He L. Effects of saline-alkaline stress on seed germination and seedling growth of Sorghum bicolor (L.) Moench. Appl Biochem Biotechnol. 2014;173(7):1680–91.
2. Zegada-Lizarazu W, Monti A. Photosynthetic response of sweet sorghum to drought and re-watering at different growth stages. Physiol Plant. 2013;149(1):56–66.
3. Zhou T, et al. Overexpression of sweet sorghum cryptochrome 1a confers hypersensitivity to blue light, abscisic acid and salinity in Arabidopsis. Plant Cell Rep. 2018;37(2):251–64.
4. Yuan X, et al. A real filed phytoremediation of multi-metals contaminated soils by selected hybrid sweet sorghum with high biomass and high accumulation ability. Chemosphere. 2019;237:124536.
5. Jiang T, et al. Integrated metabolomic and transcriptomic analysis of the anthocyanin regulatory networks in Salvia miltiorrhiza Bge flowers. BMC Plant Biology. 2020;20(1):349.
6. Gutierrez-Gamboa G, et al. Rootstock effects on grape anthocyanin, skin and seed proanthocyanidins and wine color and phenolic compounds from Vitis vinifera L Merlot grapevines. J Sci Food Agric. 2019;99(6):2846–54.
7. Chaves-Silva S, et al. Understanding the genetic regulation of anthocyanin biosynthesis in plants - Tools for breeding purple varieties of fruits and vegetables. Phytochemistry. 2018;153:11–27.
8. Lap B, Rai M. Playing with colours: genetics and regulatory mechanisms for anthocyanin pathway in cereals. Biotechnol Genet Eng Rev. 2021;37(1):1–29.
9. Choi YM, et al. Isoflavones, anthocyanins, phenolic content, and antioxidiant activities of black soybeans (Glycine max (L.) Merrill) as affected by seed weight. Sci Rep. 2020;10(1):19960.
10. Dixon R, Xie D, Sharma S. Proanthocyanidins—a final frontier in flavonoid research? New Phytol. 2005;165(1):19–28.
11. Tanner G, et al. Proanthocyanidin biosynthesis in plants Purification of legume leucoanthocyanidin reductase and molecular cloning of its cDNA. J Biol Chem. 2003;278(34):31647–56.
12. Zhou Y, et al. Metabolite Profiling of Sorghum Seeds of Different Colors from Different Sweet Sorghum Cultivars Using a Widely-Targeted Metabolomics Approach. Int J Genomics. 2020;2020:6247429.
13. Hu Y, Cai B, Huan T. Enhancing Metabolome Coverage in Data-Dependent LC-MS/MS Analysis through an Integrated Feature Extraction Strategy. Anal Chem. 2019;91(22):14433–41.
14. Chen W, et al. A novel integrated method for large-scale detection, identification, and quantification of widely targeted metabolites: application in the study of rice metabolomics. Mol Plant. 2013;6(6):1769–80.
15. Wang A, et al. A comparative metabolomics study of flavonoids in sweet potato with different flesh colors (Ipomoea batatas (L.) Lam). Food Chem. 2018;260:124–34.
16. Fraga CG, et al. Signature-discovery approach for sample matching of a nerve-agent precursor using liquid chromatography-mass spectrometry, XCMS, and chemometrics. Anal Chem. 2010;82(10):4165–73.
17. Li Y, et al. A key structural gene, AaLDX1, is involved in anthocyanin biosynthesis in all red-fleshed kiwifruit (Actinidia arguta) based on transcriptome analysis. Gene. 2018;648:31–41.
18. Li Y, et al. A Genome-wide View of Transcriptome Dynamics During Early Spike Development in Bread Wheat. Sci Rep. 2018;8(1):15338.
19. Ampomah-Dwamena C, et al. The kiwifruit lycopene beta-cyclase plays a significant role in carotenoid accumulation in fruit. J Exp Bot. 2009;60(13):3765–79.
20. Hichri I, et al. Recent advances in the transcriptional regulation of the flavonoid biosynthetic pathway. J Exp Bot. 2011;62(8):2465–83.
21. Cho K, et al. Network analysis of the metabolome and transcriptome reveals novel regulation of potato pigmentation. J Exp Bot. 2016;67(3):1519–33.
22. Vandesompele J, et al. Accurate normalization of real-time quantitative RT-PCR data by geometric averaging of multiple internal control genes. Genome Biology. 2002;3(7):RESEARCH0034.
23. Abdelrahman M, et al. Metabolome-Based Discrimination Analysis of Shallot Landraces and Bulb Onion Cultivars Associated with Differences in the Amino Acid and Flavonoid Profiles. Molecules (Basel, Switzerland). 2020;25(22):5300.
24. Jaakola L. New insights into the regulation of anthocyanin biosynthesis in fruits. Trends Plant Sci. 2013;18(9):477–83.
25. Han Y, et al. Ectopic expression of apple F3'H genes contributes to anthocyanin accumulation in the Arabidopsis tt7 mutant grown under nitrogen stress. Plant Physiol. 2010;153(2):806–20.
26. Solomon A, et al. Antioxidant activities and anthocyanin content of fresh fruits of common fig (Ficus carica L.). J Agric Food Chem. 2006;54(20):7717–23.
27. Ercisli S, et al. Color and antioxidant characteristics of some fresh fig (Ficus carica L.) genotypes from northeastern Turkey. Plant Foods Hum Nutr (Dordrecht, Netherlands). 2012;67(3):271–6.
28. Espley R, et al. Red colouration in apple fruit is due to the activity of the MYB transcription factor, MdMYB10. The Plant J. 2007;49(3):414–27.
29. Wang Z, et al. Ficus caricaRegulation of Fig (L) Fruit Color: Metabolomic and Transcriptomic Analyses of the Flavonoid Biosynthetic Pathway. Front Plant Sci. 2017;8:990.
30. Scharf K, et al. The plant heat stress transcription factor (Hsf) family: structure, function and evolution. Biochem Biophys Acta. 2012;1819(2):104–19.
31. Peng L, et al. Ecotopic over-expression of PoCHS from Paeonia ostii altered the fatty acids composition and content in Arabidopsis thaliana. Physiol Plant. 2020;172(1):64–76.
32. Qi Y, et al. Transcriptome profiling provides insights into the fruit color development of wild Lycium ruthenicum Murr. from Qinghai-Tibet Plateau. Protoplasma. 2020;307(9):110570.
33. Tai D, et al. A Malus crabapple chalcone synthase gene, McCHS, regulates red petal color and flavonoid biosynthesis. PLoS One. 2014;9(10):e110570.
34. Dick C, et al. Arctic mustard flower color polymorphism controlled by petal-specific downregulation at the threshold of the anthocyanin biosynthetic pathway. PLoS One. 2011;6(4):e18230.

Publisher’s Note
Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.