Study of differentially expressed genes related to plant height and yield in two alfalfa cultivars based on RNA-seq

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Background. Alfalfa (*Medicago sativa* L.) is a kind of forage with high relative feeding value in farming and livestock breeding, and is of great significance to the development of animal husbandry. The growth of the aboveground part of alfalfa is an important factor that limits crop yield. Clarifying the molecular mechanisms that maintain vigorous growth in alfalaf may contribute to the development of molecular breeding for this crop.

Methods. In the present study, the phenotypes of five alfalfa cultivars were evaluated. We evaluated five alfalfa cultivars for growth related traits (including WL 712, WL 525HQ, Victoria, Knight 2 and Aohan. RNA-seq was performed on the stems of both cultivars. GO enrichment analysis was conducted on all differentially expressed genes (DEGs).

Result. Among the differentially expressed genes that were up-regulated in the fast-growing cultivar, GO analysis revealed enrichment in the following seven categories: formation of water-conducting tissue in vascular plants, biosynthesis and degradation of lignin, formation of the primary or secondary cell wall, cell enlargement and plant growth, cell division and shoot initiation, stem growth and induced germination, and cell elongation. KEGG analysis showed that differentially expressed genes were annotated as being involved in plant hormone signal transduction, photosynthesis, and phenylpropanoid biosynthesis. KEGG analysis also showed that up-regulated in the fast-growing cultivar were members of the *WRKY* family of transcription factors related to plant growth and development, members of the *NAC* and *MYB* gene families related to the synthesis of cellulose and hemicellulose, and the development of secondary cell wall fibres, and finally. *MYB* family members act as activators or inhibitors and are involved in plant growth regulation. Our research results not only enrich the transcriptome database of alfalfa, but also provide valuable information for explaining the molecular mechanism of fast-growth, and can provide reference for the actual production of alfalfa at the same latitude and similar soil in the world.

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Abstract

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Key words: Medicago sativa, RNA – seq, DEGs, Stem elongation, Vigorous-growing, Slow-growing

Introduction

The stem is an important vegetative organ between the root and leaf of a plant and transports nutrients and water (Ernest et al., 2020). The stems of alfalfa also play a role in photosynthesis, nutrient storage, and regeneration (Sena, 2014). In the process of stem growth and development, stem tips grow continuously, whereas branches, leaves, and lateral branches are produced successively, which together constitute a huge branch system (Yu et al., 2015; Jaykumar & Mahendra, 2016). The degree of stem development is closely related to the life cycle of plants (Sophia et al., 2021), especially the aboveground biomass of the plant (Kleyer et al., 2019). Alfalfa, with stems and branches as the main components of biomass yield, is a typical representative crop.

Alfalfa is a feed crop with a high economic value (Kumar et al., 2018). In addition to its stress resistance properties, it has been the focus of research because of its perennial nature and high nutritional value (Wang et al., 2017; Diatta, Doohong & Jagadish, 2021). The stems and leaves of alfalfa have a high nutrient content and are the main parts areas of animal forage (Sulc et al., 2021). Owing to the cross-pollination of alfalfa, most cultivars have a complex genetic background. Restricted by its genetic characteristics, growth performance and nutritional quality are uneven (Bambang et al., 2021). Alfalfa stalks are composed of nodes and internodes, which affect plant height and yield. The height and stem diameter of alfalfa are important factors that restrict its biomass (Monirifar, 2011). Therefore, increasing the number of alfalfa vegetative branches, vegetative growth time, and delaying the flowering time of plants are crucial for improving the nutritional quality and yield of forage grass (Aung et al., 2015). Previous studies have reported significant differences in alfalfa plant height and ~~hay~~ yield among cultivars (Ziliotto et al., 2010). The ~~WL alfalfa series exhibited~~ the best growth performance when compared ~~to different alfalfa~~ cultivars (Tetteh *&* Bonsu, 1997). Plant spacing and light significantly effect ~~on~~ alfalfa forage yield and weed inhibition ~~in the field~~ (Celebi et al., 2010). Compound fertilizers can increase the nutrient content of soil and improve the yield of alfalfa (Iryna, Rudra & Doohong, 2021; Na et al., 2021). Additionally, the growth and development periods of alfalfa are equally important for its yield (Martin et al., 2010). During the growth of alfalfa, the budding stage that has excellent nutritional quality and biomass yield has always been a period of concern for breeders ~~around the world~~ (Fan et al., 2018). Currently, research on the growth performance of alfalfa mainly focuses on the physiological level. Few reports have revealed the molecular mechanism of alfalfa stem elongation and diameter enlargement ~~and its effect on biomass at the gene level~~.

Owing to the lack of a complete reference genome sequence, previous studies on the stress-response genes of alfalfa have used nonparametric transcriptome analysis (Yuan et al., 2020; Wang et al., 2021; Gao et al., 2016; Arshad, Gruber & Hannoufa et al., 2018). Reference-free transcriptome refers to the sequencing of eukaryotic transcriptomes in the absence of a reference genome. After obtaining the original data for eukaryotic nonparametric transcriptome sequencing, the quality control splicing is first performed ~~into used~~ to generate unigenes, which are then used as the reference sequence for subsequent analysis. However, with the availability of whole-genome sequencing and annotation of alfalfa (Zhongmu 1), studying the alfalfa genome has become easier (Zhang et al., 2021). Transcriptome sequencing is the study of all mRNAs present in a given sample, which is the basis for the study of gene function and is

important for understanding the development of organisms (Wang, Gerstein & Snyder, 2009). With the advantages of high-throughput, high accuracy, and high sensitivity, RNA-seq can be used to study changes in the expression level of transcripts to understand or reveal the intrinsic relationship between gene expression and biological phenotypes (Guo et al., 2021). At present, RNA-seq technology has become a common method to study the growth and development of many plants (Chen et al., 2020; Kim et al., 2021; Zheng et al., 2021). Next-generation high-throughput sequencing technology can be used to comprehensively obtain the transcript information of alfalfa and screen out the significantly different genes related to stem elongation and diameter enlargement.

The growth rate of alfalfa is an important factor of alfalfa that affects plant height and yield (Yan et al., 2021). Exploring the molecular mechanisms in alfalfa that regulate growth rate may be helpful to improve the yield. Previous studies reported that the application of gene editing technology may be more efficient than traditional techniques such as cross-breeding (Yan et al., 2021). We identified differentially expressed genes (DEGs) in the stem of alfalfa "WL 712" (USA, Fall Dormancy = 10.2) and "Aohan" (China, Fall Dormancy = 2.0) using RNA-seq, further identified the key genes regulating vigorous-growing of alfalfa by bioinformatics analysis and predicted their functions. These results may be helpful in clarifying the molecular mechanism that regulate growth rate in alfalfa, establishing a regulatory network of the growth and development of dominant cultivars, and laying a theoretical foundation for molecular breeding and the introduction of dominant cultivars.

Materials & Methods

Characterisation of phenotypic traits

Five cultivars of alfalfa, *Medicago sativa* (WL 712, Victoria, WL 525HQ, Knight 2, and Aohan) were planted at the experimental herbage station of Shihezi University, Xinjiang, China (N44 ° 20 ', E88 ° 30', altitude 420 m) (**Table S1a**). Its characteristic is temperate continental arid climate, with an average annual temperature of 8.1°C. Before planting, we adopted the “S” shaped sampling method, and nine soil samples were obtained. The nutrient status of the soil (20 cm) was as follows: available nitrogen 92.6 mg/kg, organic matter 12.4 g/kg, available potassium 168.5 mg/kg, available phosphorus 33.2 mg/kg, and pH 7.26 (**Table S1b**).

In June 2019 and 2020, alfalfa was planted in a 40 m^2 plot using a completely randomised design. To ensure consistency among the cultivars, thirty-six stems with well-growing single alfalfa were collected from each cultivar. Single-row planting method with sampling plant spacing of 40 cm and row spacing of 60 cm, with three biological replicates per cultivar. At the budding stage, agronomic traits of five randomly selected plants were determined from each of the three biological replicates. The absolute distance from the root to the top of the main stem was calculated as plant height by using a ruler. The number of branches and nodes was counted. The stem diameter and internode length were measured by using calipers. The leaf area was measured by using a leaf area meter. Five plants in each row were randomly selected and weighed, and the average value was calculated as the total fresh weight per plant. By comparing and analyzing the growth indexes of different varieties, it was finally determined that WL 712 represented a vigorous and fast-growing variety and Aohan represented a short and slow-growing variety (**Fig. 1**).

Cultivation of experimental materials and sample collection

Stems of WL 712 and Aohan were collected and cut into 8 cm pieces, leaving an axillary

bud. The stems were cultivated on cutting beds in the greenhouse (light/dark: 16 h / 8 h, Temp: 25 °C / 20 °C, humidity 70%) of the Beiyuan campus of Shihezi University for 20 days, and surviving plants were transplanted into plastic pots (diameter 32 cm, height 35 cm). Nutrient soil: vermiculite = 1: 1 (cultivation and management methods were consistent). More than 30 individual plants of both WL 712 and Aohan survived in the greenhouse. Five plants each of WL 712 and Aohan alfalfa were randomly selected and the plant height, internode length, stem diameter, leaf area and yield were determined.

At the budding stage (~~the surviving plants were transplanted and planted for about 42 days~~), stems (approximately 1.5 cm) of each cultivar were collected, quickly frozen in liquid nitrogen. Three biological replicates were used for per cultivar. WJ1, WJ2 and WJ3 represent samples from the WL 712 cultivar. AJ1, AJ2 and AJ3 represent samples from the Aohan cultivar. Finally, six samples were used for RNA-seq.

Library construction and RNA-seq

Total RNA was isolated from stems using the RNeasy Plant Mini Kit (Qiagen, Germany). A total of 3 μg RNA per sample was used to build the library. Sequencing libraries were generated using a NEBNext Ultra RNA Library Prep Kit (NEB, USA). Messenger RNA was purified from each sample using magnetic beads and fragmented with divalent cations at elevated temperature. First-strand cDNA was obtained using segmented mRNA as template and random oligonucleotide as primer. Then, the second strand of cDNA was obtained in DNA polymerase I system. The double-stranded cDNA ~~were~~ purified using AMPure XP Beads (Beckman Coulter, Beverly, USA). The double-stranded cDNA was ligated to the sequencing adaptor after terminal repair and A tail, and 250-300 bp cDNA was obtained using AMPure XP beads. Finally, the PCR system was amplified, and the PCR products were purified again using AMPure XP beads to obtain the libraries.

Library quality was examined using the Agilent Bioanalyzer 2100 system. The effective concentration of the library ($\geqslant$2 nM) was quantified using qRT-PCR. After passing the inspection, the libraries were pooled and sequenced on the Illumina HiSeq X-10 (California, USA) platform by Beijing Novo Biotech Company, Ltd. Finally, each sample contained an average of 6.63 G of valid data, and 4.42×10^7 clean reads.

Quality control

To ensure the accuracy of data analysis, we filtered the original data and examined the sequencing error rate. Using in-house Perl scripts to process the raw reads of fastq format. Removing reads containing adapters, ploy-N sequences, and low-quality from the raw data to obtain clean reads. The Q20, Q30, and GC contents of the clean data were calculated. All subsequent analyses depend on clean data, high quality.

RNA-seq data analysis

The analysis and calculation of all transcriptome data referred to previous research report (Trapnell et al., 2012). In brief, the index of the reference genome was constructed using HISAT2 v2.2.1. The paired-end clean reads were obtained using HISAT2 v2.2.1 (https://cloud.biohpc.swmed.edu/index.php/s/fE9QCsX3NH4QwBi/download) aligned to the reference genome Zhongmu No. 1 (https://figshare.com/articles/dataset/genome_fasta_sequence_and_annotation_files/12327602) to obtain mapped reads (Mortazavi, Williams & McCue, 2008). We also analysed the proportion of mapped reads in the exons, introns, and intergenic regions of the genome.

The clean reads aligned to Zhongmu No. 1 were quantified using FeatureCounts v1.5.0-p3. Gene expression was tested by FPKM (fragments per kilobase of transcript per million fragments mapped), and differences between WL 712 and Aohan FPKM values were compared using FeatureCounts v1.5.0-p3.

Differential expression analysis of the two comparison combinations was performed using the DESeq2 R package (1.16.1) (https://www.bioconductor.org/packages/release/bioc/html/DESeq2.html). DESeq2 determines the differential expression in digital gene expression data using a model based on a negative binomial distribution. The corrected P-values and $|\log_2 \text{foldchange}|$ are thresholds for significant differential expression. P-values were adjusted using the Benjamini & Hochberg method.

Gene Ontology (GO) (http://www.geneontology.org/) enrichment and KEGG (Kyoto Encyclopedia of Genes and Genome) (http://www.genome.jp/kegg/) statistical analysis of DEGs were performed using the clusterProfiler R package. A corrected P-value less than 0.05 was used as the threshold for significant enrichment of differentially expressed genes.

qRT-PCR

The accuracy of the RNA-seq was verified by qRT-PCR. Total RNA ~~were~~ isolated from stems, and cDNA was synthesised by using the PrimeScript RT reagent Kit (Takara, Tokyo, Japan). Alfalfa *β-Actin 2* was used as the internal gene. The primers in **Table S9** were used for qRT-PCR. qRT-PCR ~~were~~ completed using the LightCycler 96/LightCycler480 system. The solution of the 20 μL system contained 0.4 μL forward primer, 0.4 μL ~~R~~everse ~~P~~rimer, 10 μL TB Green Fast qPCR Mix (2X) (Takara, Tokyo, Japan) and 2 ng cDNA. The PCR procedure included 45 cycles, with 3 technical repetites for each reaction. According to Kenneth report, the relative expression of each gene was calculated (Livak & Schmittgen, 2001).

Statistical Analysis

All statistical analysis was using *SPSS* software (version 17; IBM Inc, USA). The data were compared using Student's t-test, and $P < 0.05$ was considered statistically significant. The power of our samples was calculated using RNASeqPower (https://bioconductor.org/packages/release/bioc/html/RNASeqPower.html), and the RNASeqpower was 94.2%.

Results

Phenotypic analysis of five alfalfa varieties

To compare the differences in the growth patterns of the five cultivars (**Table S1a**), plant height, internode length and stem diameter of alfalfa at different growth stages were continually measured in 2019 and 2020 (**Fig. 2, Table S10**). There were no significant differences in plant height, internode length or stem diameter among cultivars at the seedling transplant stage. After the budding stage, plant height, internode length and stem diameter of different alfalfa varieties reached a plateau and remained relatively stable (**Fig. 2a-c**). In 2019 and 2020, WL 712 and Aohan represented tall and short phenotypes, respectively (**Fig. 2d**). Comparing the agronomic traits of alfalfa at the budding stage in 2019 and 2020, the plant height of WL 712 was approximately 1.78 and 1.91 times those of Aohan, respectively, and the stem diameter of WL 712 was approximately 1.90 and 1.92 times those of Aohan (**Fig. 2d-e**). The internode length and number of lateral branches in WL 712 were significantly larger than those in Aohan ($P < 0.01$), whereas the number of main branches in WL 712 was significantly lower ($P < 0.05$) (**Fig. 2f, Fig. 3a-b**).

To identify the correlation between internode length and stem diameter and other traits, the fresh weight, leaf-stem ratio, and dry weight of the five cultivars were also determined. The results showed that the production performances of WL 712 and Aohan were significantly different ($P < 0.05$) (**Fig. 3c-f**). Phenotypic correlation analysis based on 8 agronomic traits was done. We found that fresh and dry weight were positively and strongly correlated with the number of lateral branches, plant height, stem diameter, and internode length, and plant height was significantly positively correlated with internode length ($P < 0.01$). In addition, the number of main branches was negatively correlated with plant height, stem diameter, and internode length ($P < 0.01$) (**Table 1**).

From the screening of five alfalfa cultivars, WL 712 and Aohan were identified as the cultivars with the most significant difference in growth performance (**Fig. 1**). The growth trend of the two varieties in greenhouse is similar to that in field. The plant height, internode length, yield per plant, leaf area and stem diameter of WL 712 alfalfa were significantly higher than those of Aohan alfalfa (**Table 2**).

Based on the above results, WL 712 and Aohan were used as the vigorous-growing and slow-growing experimental cultivars, while the stem base tissue with the budding stage (the surviving plants were transplanted and planted for about 42 days) was used for RNA-seq.

RNA-seq analysis

Using RNA-seq, we obtained 2.74×10^8 raw reads. The sequence error rate of a single base position was 0.03%, and the average GC content was 41.65%. After filtering from the raw data, 2.65×10^8 (96.94%) clean reads (39.76 G) were obtained. The phred values were greater than 97% and 93% at Q_{20} and Q_{30}, respectively (**Table S1c**). The pearson coefficient showed that the homology among the samples within the group was higher than 84.6% (**Fig. S1**).

We aligned the clean reads with the reference genome. The average proportions of exons, introns and intergenic regions in AJ samples were 72.72%, 3.61%, and 23.67%, respectively. Similarly, the WJ samples accounted for 74.14%, 2.96%, and 22.90%, respectively (**Table S2**). The reads aligned to the intron region may have been derived from the precursor mRNA. The reads aligned to the intergenic region may have been derived from ncRNAs.

Additionally, according to the comparison of RNA-seq data from WL 712 and Aohan, the

RNASeqpower of our sample was 94.2%. The result may be beneficial to screen and explore the functional DEGs related to the vigorous-growing of alfalfa. These results demonstrated that the experiments were reproducible and that the data were accurate.

Identification and functional annotation of DEGs in WL 712 and Aohan

Generally, the gene expression value of RNA-seq is evaluated as fragments per kilobase of transcript per million mapped reads (FPKM), which corrects the sequencing depth and gene length successively (**Fig. S2**). More than 90% of the clean reads were successfully mapped to the alfalfa genome. To clarify the function of the DEGs ~~in~~ WL 712 and Aohan, we performed GO and KEGG enrichment analyses. In total, 954 DEGs were significantly enriched and assigned to 35 GO terms. Compared to Aohan, WL 712 upregulated 578 genes and downregulated 376 genes. Among the molecular function, "*protein heterodimerization activity*" [GO:0046982] (114 DEGs, 11.95%) was the highest proportion, followed by "*UDP-glycosyltransferase activity*" [GO:0008194] (99 DEGs, 1.04%) and "*translation factor activity, RNA binding*" [GO:0008135] (86 DEGs, 9.01%). Among the cell components, "*bounding membrane of organelle*" [Go:0098588] (57 DEGs, 5.97%) represented the largest cluster, followed by "*whole membrane*" [Go:0098805] (49 DEGs, 5.13%) and "*peptidase complex*" [Go:1905368] (44 DEGs, 4.61%). Among the biological processes, "*translational elongation*" [GO:0006414] (41 DEGs, 4.30%) represented the largest cluster (**Table 3, Table S3, Fig. 4**).

Based on biological system network, the function of DEG was identified using KEGG classification. A total of 1324 genes were enriched in 110 KEGG pathways (**Fig. 5**). "*Carbon metabolism*" [ath01200] (103 DEGs, 7.8%) and "*Ribosome*" [ath03010] (96 DEGs, 7.3%) were the most abundant pathways; followed by "*Biosynthesis of amino acids*" [ath01230] (81 DEGs, 6.1%), "*RNA transport*" [ath03013] (54 DEGs, 4.1%), "*Plant-pathogen interaction*" [ath04626] (52 DEGs, 3.9%), "*Protein processing in endoplasmic reticulum*" [ath04141] (52 DEGs, 3.9%) and "*Plant hormone signal transduction*" [ath04075] (44 DEGs, 3.2%) (**Table S4**).

Expression and regulation of DEGs in WL 712 and Aohan

KEGG analysis showed that DEGs related to stem elongation and diameter enlargement were widely involved in biological processes such as hormone signalling, photosynthesis and transcriptional regulation (**Table S5**).

Plant hormone signal transduction (Ath04075) involves many hormones that regulate the growth and development, such as auxins, cytokinines, gibberellins, brassinosteroids, jasmonic acid, and ethylene. Twelve DEGs were enriched in the auxin-mediated signalling pathway, including *auxin-responsive protein SAUR* (*SAUR*), *auxin-induced protein X10A* (new gene) and *auxin transporter-like protein* (*LAX*). Among these, *IAA9*, *IAA6*, *SAUR50*, *SAUR32* and *SAUR36* were significantly upregulated. In the cytokinin-mediated signalling pathway, four DEGs were enzyme genes, such as *adenylate isopentenyltransferase 5* (*IPT5*), *7-deoxyloganetin glucosyltransferase* (*UGT85A24*), *cytokinin dehydrogenase 6* (*CKX6*), and *cytokinin hydroxylase* (*CYP735A2*). *DELLA protein GAI* (*GAI*), *f-box protein GID2* (*GID2*), and *transcription factor PIF4* (*PIF4*) were enriched in the gibberellin-mediated signalling pathway. *Serine/threonine-protein kinase BSK8* (*BSK8*), *serine/threonine-protein kinase BSK1* (*BSK1*), and *Cyclin-D3-3 (CYCD3-3)* were enriched in the brassinosteroid-mediated signalling pathway. Five DEGs were enriched in the jasmonic mediated signalling pathway, including *Coronatine-insensitive protein homolog 1a* (*COII A*), *Protein TIFY 6 B* (*TIFY6B*), *Protein TIFY 11 B* (*TIFY11B*), *Protein TIFY 10 B* (*TIFY10B*) and *Protein TIFY 3 B* (*TIFY3B*). Four

upregulated DEGs were enriched in the ethylene-mediated signalling pathway, including *ethylene receptor* (*ETR1*), *mitogen-activated protein kinase kinase 4* (*MKK4*), *mitogen-activated protein kinase homolog MMK1* (*MMK1*), and *protein ethylene insensitive 3* (*EIN3*).

Fifteen DEGs were enriched in the photosynthetic ~~(ath00195)~~ pathway. Among them, *PPL1*, *PETC*, *PSBR*, *PSBS*, *PSAG*, *PSAO*, *PSB 27* and *PSB 28* were related to the photoreaction. *PLSN 2* was related to the activity of the chloroplast NAD(P)H dehydrogenase (NDH) complex. *ATPF 2* and *ATPC* are related to ATPase activity. Additionally, two oxygen-evolving enhancer proteins and ferredoxins have been identified. In the photosynthesis-antenna protein (ath00196) pathway, eleven DEGs were classified into *chlorophyll a-b binding proteins* and *chlorophyll a/b binding proteins*, which were expressed in chloroplasts. In the MAPK signalling (ath04016) pathway, twenty-two DEGs were mainly involved in biotic stress (pathogen infection), abiotic stress (cold/salt/drought/osmotic stress), and hormone synthesis during root growth and wounding responses.

Furthermore, the TCA cycle (ath00020), carbon fixation in photosynthetic organisms (ath00710), glycolysis/gluconeogenesis (ath00010), ribosome (ath03010), amino sugar and nucleotide sugar metabolism (ath00520), pyruvate metabolism (ath00620), and phenylpropanoid biosynthesis (ath00940) ~~were also~~ closely related to alfalfa growth (**Table S5**). In the TCA cycle pathway, 12 DEGs ~~played~~ a role in catalysis of the pyruvate dehydrogenase complex. In addition, *ATP-citrate synthase alpha chain protein 1* (*ACLA1*) and *2 malate dehydrogenases* (*MDH*) were identified. *Pyrophosphate--fructose 6-phosphate 1-phosphotransferase subunit beta* (*PFP*) and *glycoaldehyde-3-phosphate dehydrogenase* (*GAPC1*) ~~were highly expressed in the~~ glycolysis/gluconeogenesis pathway. Seven *glyceraldehyde-3-phosphate dehydrogenases* (*GAPDH*) were enriched in carbon fixation in the photosynthetic organism pathway and were highly expressed ~~in the cytoplasm or chloroplasts~~. Ribosomal proteins predominated in the ribosomal pathway and included 30 s (*RPS1*, *RPS13*, *RPSQ*, *RPS16*), 40s (*RP24a*, *RP30a*, *RP15d*, *RP10a*, *RP20a*), 50s (*RPL28*, *RPMJ*, *RPL31*, *RPLX*) and 60s (*RPP3a*, *RPL21e*, *RPL37a*, *RPL37B*). *Dihydrolipoyllysine-residue acetyltransferase component 2 of the pyruvate dehydrogenase complex* (*At3g13930*) and *malate dehydrogenase* (*mMDH*) were highly expressed ~~in~~ the pyruvate metabolic pathway. *Beta-glucosidase 44* (*BGLU44*), *beta-amylase 1* (*BAM1*), *acid beta-fructofuranosidase* (*VCINV*), and *probable fructokinase-4* (*At3g59480*) were highly expressed in the starch and sucrose metabolism pathways. The genes ~~with high expression~~ in the phenylpropanoid biosynthesis pathway were *Probable cinnamyl alcohol dehydrogenase* (*CAD2*), *beta-glucosidase 46* (*BGLU46*), *trans-cinnamate 4-monooxygenase* (*CYP73A3*), and *3 peroxidases* (*PER*).

DEGs enriched in a variety of biological processes

All DEGs were analysed using GO and KEGG analyses. We found that seven groups of DEGs plausibly related to stem elongation and diameter expansion, including formation of water-conducting tissue in vascular plants, cell division and shoot initiation, biosynthesis and degradation of lignin, cell enlargement and plant growth, formation of the primary or secondary cell wall, cell elongation, and stem growth and induced germination (**Table S6**). *Eukaryotic translation initiation factor 5A-1* (*EIF5A*), *mitogen-activated protein kinase kinase kinase 3* (*ANP3*), and *alpha, alpha-trehalose-phosphate synthase* (*TPS6*) ~~were~~ involved in the formation of water-conducting tissues **(Fig. 7d)**. Fourteen DEGs were enriched in lignin biosynthesis and degradation. Peroxidases play an important role in this process. Additionally, *peroxidase 47*

(*PER47*) is a novel gene **(Fig. 6a)**. Eleven DEGs were enriched in the formation of the primary or secondary cell wall. *Cellulose synthase A catalytic subunit* (*CESA*) ~~plays an active role~~ **(Fig. 6b)**. Eighteen DEGs were enriched in cell enlargement and plant growth. AUXs, such as *auxin-responsive protein* (*IAA9*), *auxin-induced protein* (*IAA6*) and *auxin transporter-like protein* (*LAX5*), ~~play an active role~~. Additionally, *auxin-induced protein X10A* is a novel gene **(Fig. 6c)**. Five DEGs were enriched in cell division and shoot initiation. Enzyme genes such as *7-deoxyloganetin glucosyltransferase* (*UGT85A24*), *cytokinin hydroxylase* (*CYP735A2*) and *cytokinin dehydrogenase 6* (*CKX6*) ~~play a dominant role~~ **(Fig. 7a)**. Two DEGs were enriched in stem growth and induced germination. Interestingly, *DELLA protein* (*GAI*) negatively regulated the gibberellin signalling pathway, whereas *F-Box protein* (*GID2*) ~~positively regulated the gibberellin signalling pathway~~ **(Fig. 7b)**. *Serine/threonine-protein kinase*（*BSK 1*）and *BSK8* are related to cell elongation **(Fig. 7c)**. Additionally, we identified genes that regulate senescence, including *protein ethylene insensitive 3* (*EIN3*) **(Fig. 6d)**. Importantly, compared with Aohan, *cellulose synthase A catalytic subunit 8* (*CESA8*), *beta-1,4-xylosyltransferase* (*IRX 9*), *probable beta-1,4-xylosyltransferase* (*IRX 14H*), *auxin-responsive protein* (*SAUR36*), *peroxidase 16* (*PER16*), and *peroxidase 51* (*PER51*) were upregulated more than 8-fold in WL 712, whereas *mitogen-activated protein kinase 3* (*MPK3*), *pathogenesis-related protein* (*At2g14610*), *peroxidase 55* (*POD55*), *beta-glucosidase 46* (*BGLU46*), and *peroxidase 15* (*POD15*) were downregulated more than 15-fold in WL 712 (**Table S7**). All the genes might related to stem growth and development were clustered together, as shown in **Fig. 6 and Fig. 7**.

Transcription factors involved in alfalfa growth and development

Transcription factors are essential in plant growth and development as protein molecules that regulate gene expression. In this study, 20 transcription factors were ~~involved in the development of alfalfa~~ (**Fig. 8a, Table S8**). Seven DEGs were upregulated, including *NAC domain-containing protein 73* (*NAC073*), *NAC domain-containing protein 10* (*NAC010*), *transcription factor MYB 46* (*MYB46*), and *NAP-related protein 2* (*NRP2*). Additionally, *WRKY transcription factor 22* (*WRKY22*), *transcription factor TGA 1* (*TGA1*) and ~~T~~*ranscription factor MYB86* were novel genes. GO annotations ~~revealed~~ that *NAC073* and *NAC010* ~~were~~ involved in the synthesis of cellulose and hemicellulose and the development of secondary cell wall fibres. Thirteen DEGs were downregulated, and the *WRKY* and *MYB* family members ~~played a dominant role~~. GO classification ~~shows~~ that *WRKY51* ~~was~~ involved in the positive regulation of salicylic acid-mediated signal transduction and negative regulation of jasmonic acid-mediated signal transduction in the defense response. *WRKY54* is a negative regulator of plant growth and development. *MYB46* is involved in secondary wall cellulose ~~bio~~synthesis as a transcriptional activator. *MYB86* is involved in lignin synthesis and accumulation. Additionally, *MYB2* inhibit~~ed~~ the expression of light-harvesting genes. All identified transcription factors were validated using qRT-PCR (**Fig. 8b**). The relative expression of *NAC081* was significantly upregulated in WL 712 plant ($P < 0.001$). The relative expression levels of most transcription factors were similar to the FPKM trend.

The reliability of RNA-seq was verified using qRT-PCR

To determine the accuracy and rationality of the data, we arbitrarily selected 11 DEGs for qRT-PCR validation. DEGs were mainly related to the formation of the primary or secondary cell wall, cell enlargement and plant growth, and ~~bio~~synthesis and degradation of lignin. The

changes in transcript abundance are shown in **Fig. 9a**. qRT-PCR revealed that *IRX9*, *CESA8*, *CESA7*, *MKK4*, *PER16*, and *PER51* were significantly upregulated in WL 712 plant ($P < 0.05$). *MPK3*, *At2g14610*, *BGLU46*, and *POD15* were significantly downregulated in WL 712 ($P < 0.05$) (**Fig. 9b**). However, the relative expression of *CAD2* between the two varieties ~~were~~ not significantly different ($P > 0.05$) and ~~were~~ inconsistent ~~according to~~ RNA-seq transcript abundance. This may have been caused by RNA-seq errors in the acceptable range. Overall, the relative expression trend of the DEGs was similar to the RNA-seq.

Discussion

Alfalfa is an important component of feed, and the growth performance of its aboveground part affect the biomass yield. The *FmS6K* gene plays an important role in regulating the development of plant stems (Sun et al., 2018). The yield of elephant grass has a strong positive correlation with internode length (Yan et al., 2021). However, the molecular regulatory mechanisms underlying the growth rate of stems and branches in alfalfa remain unclear. In this study, the growth difference between the tall and fast-growing variety WL 712 and the short and slow-growing variety Aohan was studied. ~~The phenotypes and RNA-seq of these two varieties were analyzed by using stems.~~ The transcriptome of those two varieties was analyzed by RNA-seq, with RNA obtained from the stem base. The difference between qRT-PCR and RNA-seq of individual DEG may be caused by the error of RNA-seq within the acceptable range. Overall, the RNA-seq data could be used for subsequent analysis. All DEGs were associated with at least one GO term; 954 significant DEGs were obtained, and seven DEG clusters were speculated to be involved in promoting fast growth (**Fig. 6, Fig.7**). Additionally, KEGG revealed that hormone signal transduction, photosynthesis and ~~phenylpropane~~ biosynthesis genes are up-regulated in the faster growing cultivar. RNA-seq also identified several novel DEGs associated with the fast growing cultivar, including *PER47* and *TIFY10A*.

Plant organ growth is influenced by both developmental processes and environmental factors (Sun et al., 2018). In many cases, these changes are due to hormone-mediated action (Verma, Ravindran *&* Kumar, 2016). In this study, auxin, cytokinin, gibberellin, ethylene, brassinosteroid, and jasmonic acidwere all implicated because their downstream targets were found among DEGs, such as *SAUR50*, *CKX6*, *GID 2*, and *GAI*. These DEGs might play a role in promoting fast growth in alfalfa. Previous studies have identified *SAURs* as a class of hormones that regulate plant growth and development and promote cell enlargement (Ren *&* Gray, 2015). Cytokinin synthesis is required to activate shoot division in apple trees with the top removed (Tan et al., 2018). Relevant studies have shown that gibberellin regulates plant organ elongation and development (Nagel, 2020). *GAI* is an inhibitor of highly conserved gibberellin signalling in plants. The *SCF* (*GID2*) complex mediates degradation of DELLA proteins (*RLG2*, *RGA*, and *GAI*), and activates and positively regulates the gibberellin signalling pathway (Dill et al., 2004). In addition, in the plant hormone signal transduction pathway, the production of hormones that play a mediating role depends on the metabolism of amino acids or fatty acids. Tryptophan in plants is not only involved in the synthesis of proteins but also the precursor of many metabolites (such as auxin) (Manol *&* Nemoto, 2012). Jasmonic acid induces plants to prioritise defense over growth by interfering with the gibberellin signalling cascade, which is usually accompanied by significant growth inhibition (Yang et al., 2012). *TIFY*, which encodes jasmonic acid repressor, was significantly upregulated in Aohan (**Table S5**). This may

explain why theAohan alfalfa is a dwarf plant.

Photosynthesis is an essential metabolic process. Twenty-nine DEGs were related to photosynthesis. For example, *PIF 1* and *PIF 3* were significantly downregulated in WL 712 (**Table S5**). These genes may play a regulatory role in the process of plant height and internode elongation. Plant height and leaf area of transgenic soybean are decreased by overexpressing *PIF4* (Arya, Singh & Bhalla, 2021). The deletion of *PIF1* and *PIF3* results in an increase in plant height, longer internodes, and late flowering (Hoang et al., 2021). The light-harvesting complex II (LHC II) functions as a light receptor and is related to the absorption of light (Gu et al., 2017; Sen et al., 2021). The up-regulation of these DEGs may enhance the photosynthesis of WL 712 and promote the growth of plants. Additionally, circadian rhythm is also involved in the regulation of plant growth and development (Venkat & Muneer, 2022). Our research found that DEGs enriched in circadian rhythm pathway were mainly related to photoperiod flowering response. (**Table S5**).

~~Driven by differences in plant tissue growth, cells continue to divide, proliferate and differentiate, eventually forming various functional organs (Huang et al., 2018; Hilde & Nathalie, 2020).~~ RNA-seq analysis ~~showed that~~ 1531 DEGs related to rape stem growth (Yuan et al., 2019). Combined analysis of proteome and RNA–seq ~~showed~~ that DEGs and DEPs of *Mikania micrantha* stems were significantly enriched in photosynthesis, carbon sequestration and plant hormone signal transduction pathways (Can et al., 2021). We identified seven DEG clusters that were involved in stem elongation and enlargement. Fourteen DEGs were ~~enriched in genes~~ annotated as being involved in lignin ~~bio~~synthesis and degradation ~~(**Fig. 6a**)~~ and ~~mostly~~ peroxidases. These genes may regulate lignin ~~bio~~synthesis and degradation in stems. The oxidation activity of peroxidases is ~~limited to the lignified region during plant development~~ (Hoffmann et al., 2020). Eleven DEGs were apparently involved the formation of the primary or secondary cell wall (**Fig. 6b**), with good representation from cellulose synthase. Previous studies reported that *CESA 4* and *CESA8* ~~were~~ specifically enriched and expressed in the stem tissue during the ~~fibre~~ development stage (Guo et al., 2021). Eighteen DEGs were enriched in the category of cell enlargement and plant growth (**Fig. 6c**) frequently involving auxin. Five DEGs were apparently involved in cell division and shoot initiation (**Fig. 7a**). Two DEGs were enriched in the categories of stem growth and induced germination (**Fig. 7b**), mainly components of the gibberellin signalling pathway. Two DEGs were potentially involved in cell elongation (**Fig. 7c**). These DEGs might play a role in stem internode elongation, diameter enlargement and lateral branch formation. Previous studies reported that *AtTPS6* completely compensates for the defects in reduced trichome and stem branching due to *csp-1* deficiency in ~~Arabidopsis~~ (Chary et al., 2008). Deletion of *IAA17* in tomatoes showed that the increase in fruit size is related to the higher ploidy level of peel cells (Su et al., 2015). Finally, the *TIFY* homologous possibly ~~related to alfalfa~~ senescence were also identified (**Fig. 6d**). In addition, we identified several members of *SPL* family, such as *SPL1*, *SPL6* and *SPL7*, which may be involved in the lateral branch development of alfalfa. Previous studies reported that *SPL13* regulates shoot branching in alfalfa (Gao et al., 2018). Overall, these DEGs may be involved in alfalfa growth and development.

Transcription factors are essential in the regulation of development, morphogenesis and environmental stress. Previous research found that most members of *NAC*, *WRKY* and *MYB* families are involved in the synthesis of lignin, cellulose, and hemicellulose (Wang et al., 2016).

The *NAC*-mediated transcription network synergistically regulates ~~bio~~synthesis of the plant secondary wall (Ryan, Zhong & Ye, 2011). *WRKY6* and *WRKY33* positively regulated ~~ABA~~ signal transduction during early development of ~~*Arabidopsis*~~ *thaliana* (Huang et al., 2016). *WRKY54* is a negative regulator of salicylic acid ~~bio~~synthesis (Li, Zhong & Palva, 2017) and can significantly increase stem diameter, leaf area, and total dry weight of plants (Amin et al., 2013). Overexpression of *AtMYB44* in tomatoes result~~ed~~ in slow growth (Shim et al., 2012). *MYB3R1* is a transcriptional repressor that regulates organ growth, and restricts plant growth and development by binding to target genes and promoters of specific genes (Wang et al., 2018). Under reduced light intensity, *MYB2* and *MYR1* act~~ed~~ as inhibitors of flowering and organ elongation, respectively (Zhao et al., 2011). ~~In this study,~~ excluding *WRKY22*, all *WRKY* members were significantly upregulated in dwarf alfalfa. Therefore, *WRKY22* may positively regulate the growth and development of WL 712. *NACs* are involved in the development of plant secondary cell walls. Among these, *NAC081* functions as a positive regulator. *MYB46* and *MYB86* might positively regulate the synthesis of cellulose and lignin, and *MYB44*, *MYB3R1* and *MYB2* might act as transcriptional repressors (**Table S8**).

Conclusion

Plant height is an important factor in determining forage biomass. The molecular characteristics of the DEGs between fast and slow growing alfalfa cultivars were identified using RNA–seq. The trend of our qRT-PCR was largely consistent with those of RNA-seq, which indicated that the RNA-seq data could be used for subsequent analysis. All DEGs were analysed using GO terms, and 954 significant DEGs were identified. KEGG analysis indicated that hormone signal transduction, ~~phenylpropane~~ biosynthesis, and photosynthesis are well represented in the fast growing cultivar. GO analysis highlighted the following seven clusters of DEGs: formation of water-conducting tissue, cell division and shoot initiation, synthesis and degradation of lignin, stem growth, formation of the primary or secondary cell wall, cell enlargement and plant growth, and induced germination and cell elongation. Additionally, the transcription factors implicated in stem elongation and diameter expansion are mainly *WRKY*, *NAC*, and *MYB* family members. In summary, our research results not only enrich the transcriptome database of alfalfa, but also provide valuable information for explaining the molecular mechanism of fast growth, and can provide reference for the ~~actual~~ production of alfalfa ~~at the same latitude and similar soil in~~ the world.

Competing Interests

The authors declare there are no competing interests.

Author Contributions

Qi Jiangjiao conceived and designed the experiments, performed the experiments, analyzed the data, prepared ~~Fig.s and/or tables,~~ and authored or reviewed drafts of the paper.

Yuxue, Wang Xuzhe and Zhang Fanfan performed the experiments.

Ma Chunhui conceived and designed the experiments, performed the experiments, analyzed the data, authored or reviewed drafts of the paper, and approved the final draft.

Availability of data and materials

The data is available at the Sequence Read Archive (SRA) of NCBI: https://www.ncbi.nlm.nih.gov/sra/?term=PRJNA807394.

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Figure 1

Slow-growing Aohan (AJ) and vigorous-growing WL 712 (WJ) plant ~~in~~ bud stage

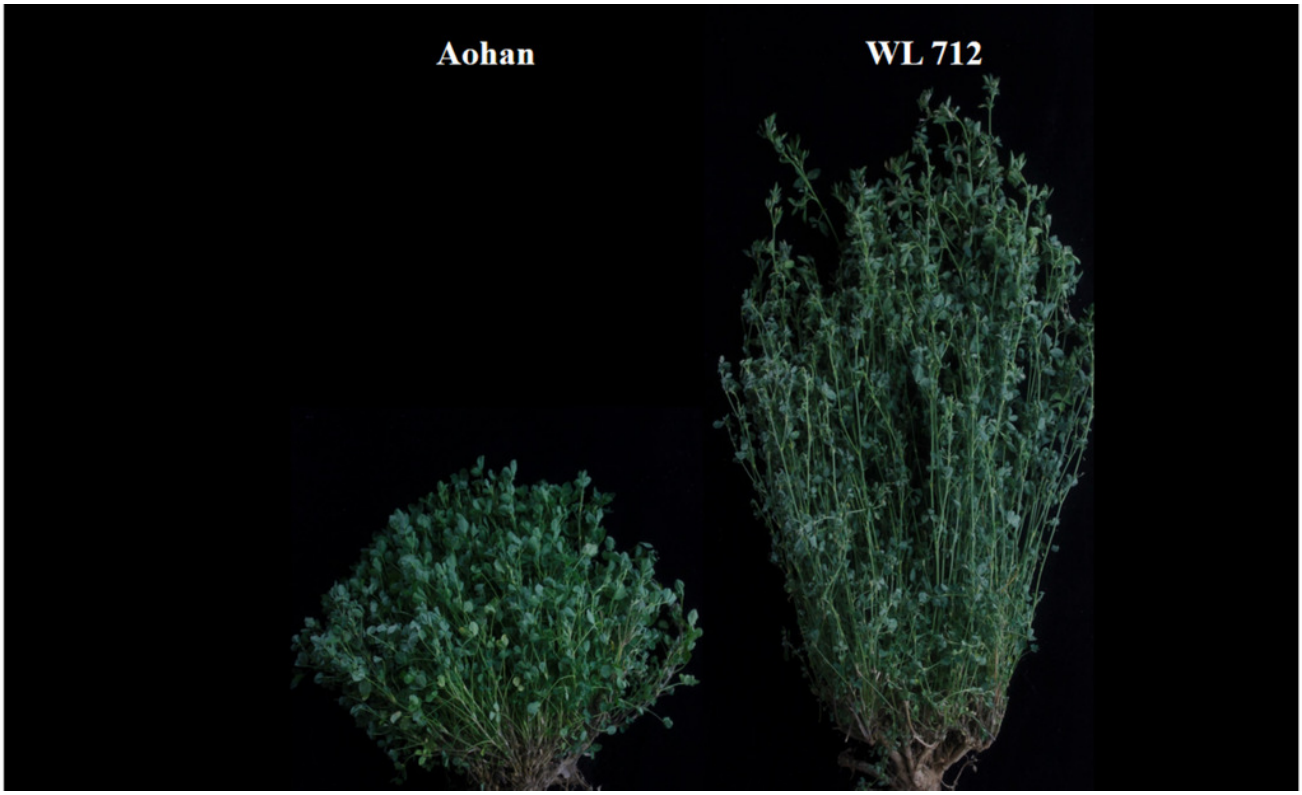

Figure 2

Phenotypic evaluation of five alfalfa cultivars.

The dynamics of plant height (a), stem diameter (b) and internode length (c) ~~development~~ of five alfalfa cultivars ~~at transplanting stage, branching stage, budding stage, early flower stage and full flower stage~~. Average plant height (d), stem diameter (e), internode length (f) of five alfalfa cultivars. The values are the average of fifteen biological replicates and error bars represent the standard deviation. Different letters indicate significant difference at $P < 0.05$ among the five cultivars as determined by Student's t test.

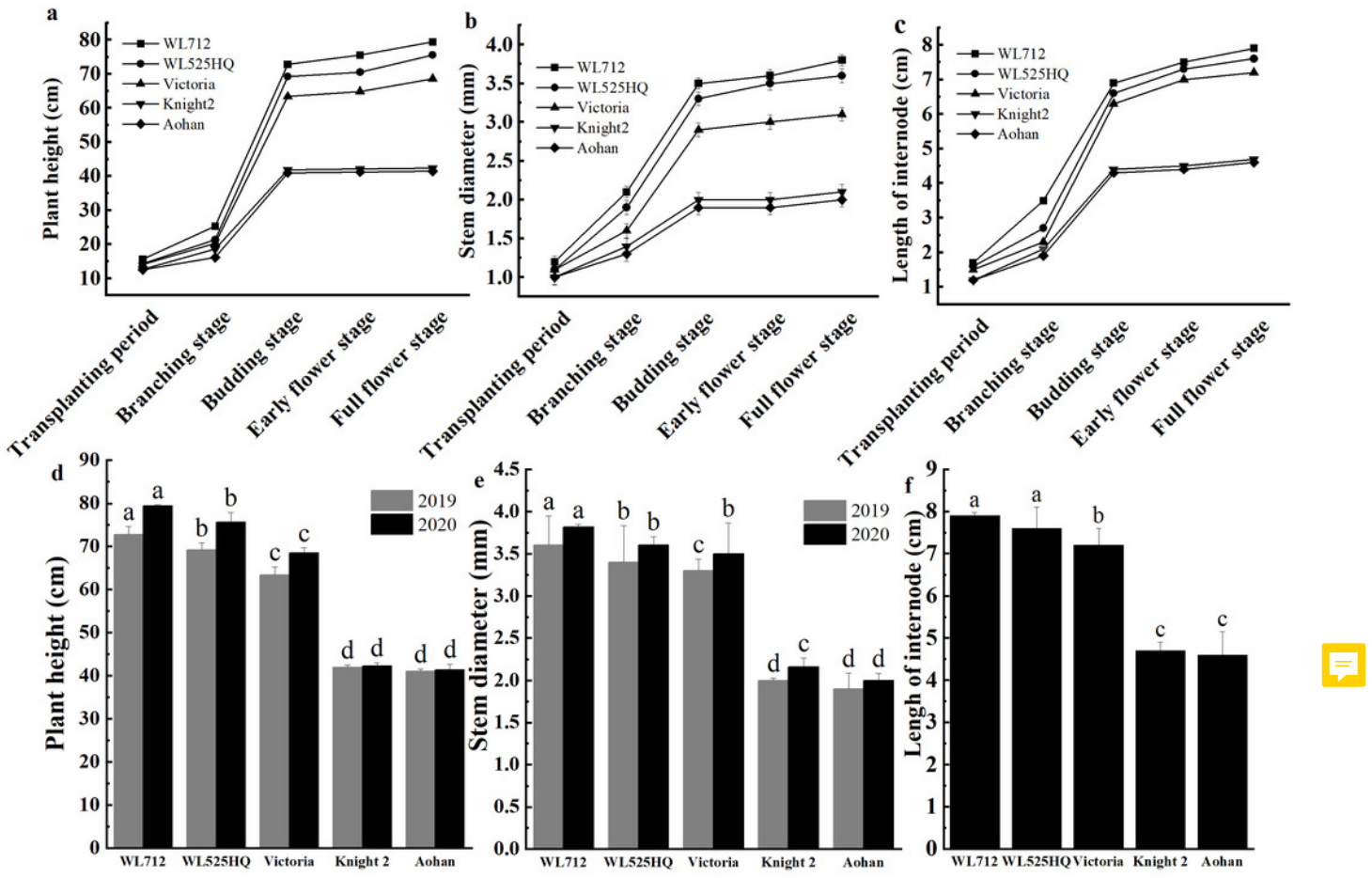

Figure 3

Phenotypic evaluation and index determination of five alfalfa cultivars ~~in bud stage~~ (42 d).

Lateral branch~~es~~ number (a), branch~~es~~ number (b), leaf area (c), fresh weight (d), leaf to stem ratio (e) and dry weight (f) of five alfalfa cultivars. The values are the average of fifteen biological replicates and error bars represent the standard deviation. Different letters indicate significant difference at $P < 0.05$ among the five cultivars as determined by Student's t test.

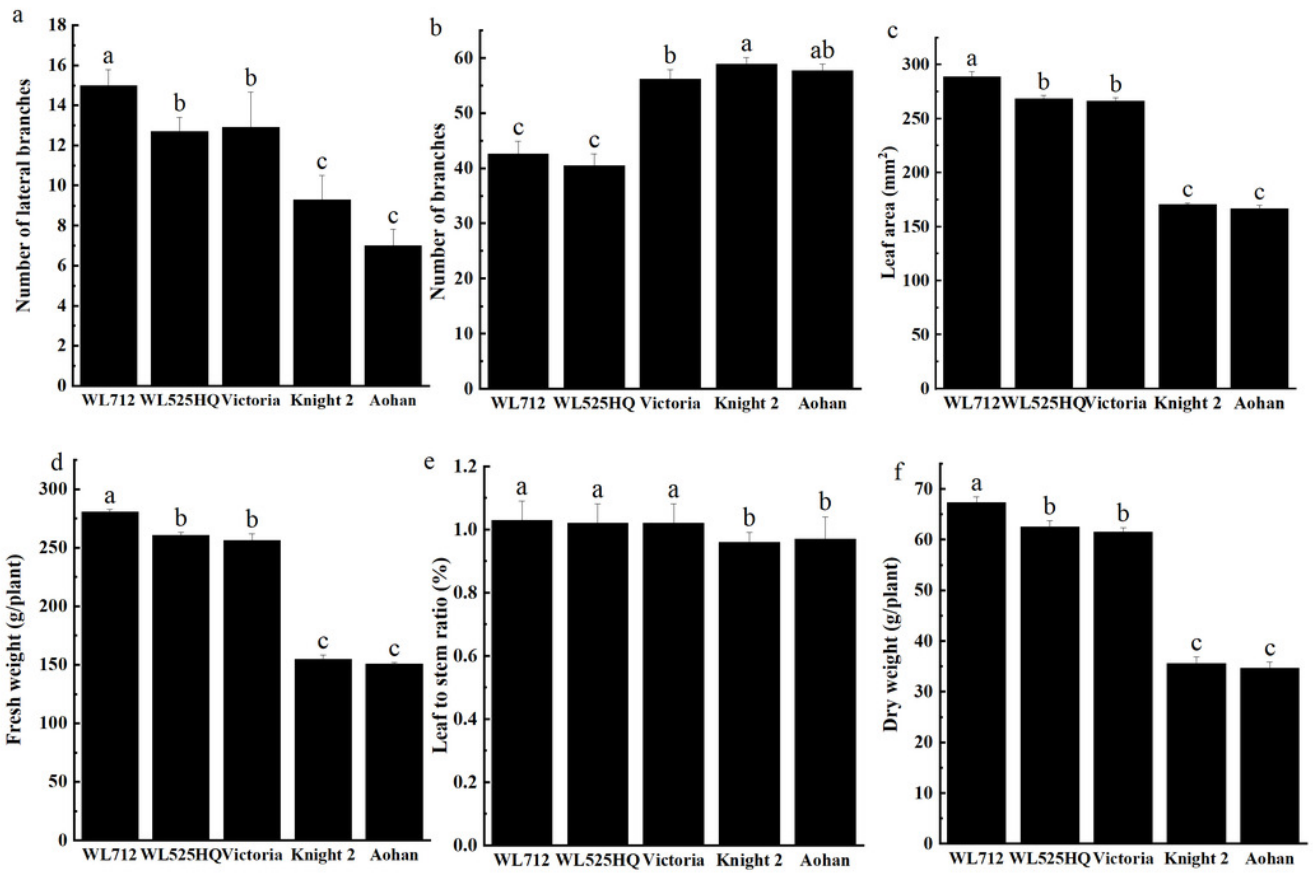

Figure 4

Scatter diagram of enriched GO functional categories.

The “GeneRatio” shows the ratio of the number of DEGs in the given category to the total number of differentially expressed genes. The size of the spot indicates the approximate number of DEGs in the category, all the spots indicate the significance level, $P < 0.05$.

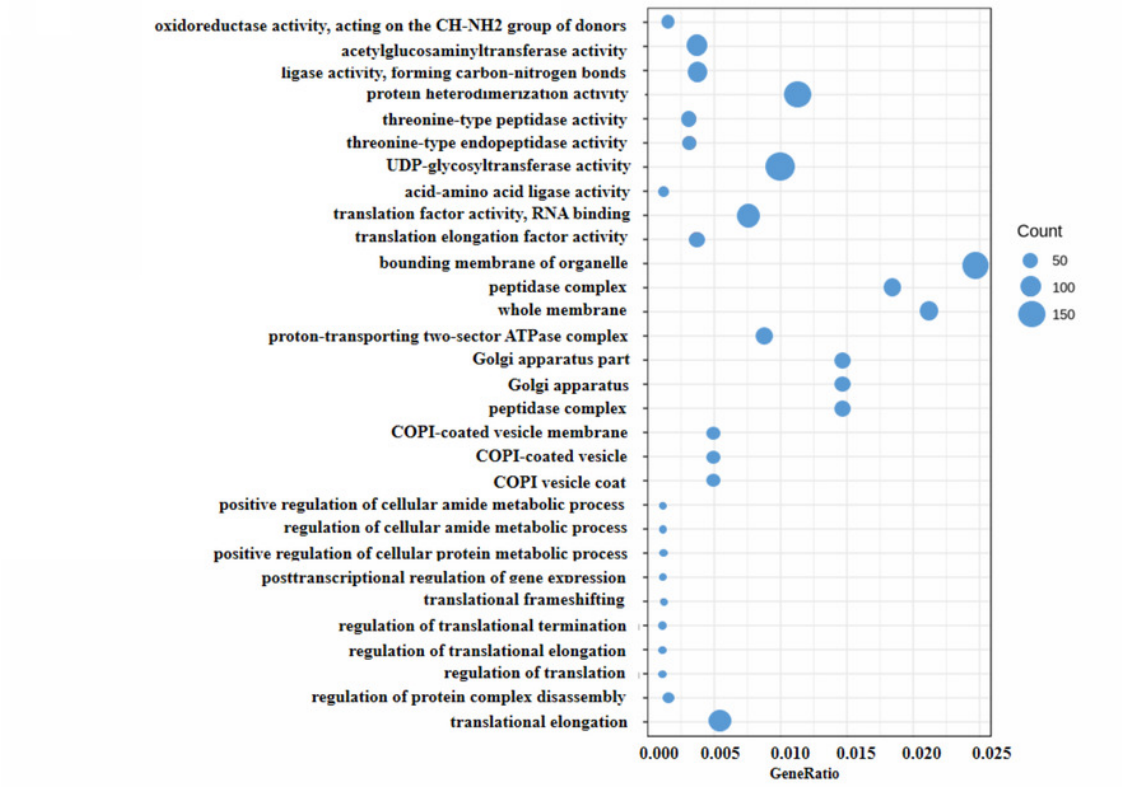

Figure 5

KEGG classification of differentially expressed genes (DEGs).

X-axis is the number of gene annotations; Y axis is the type of KEGG pathway.

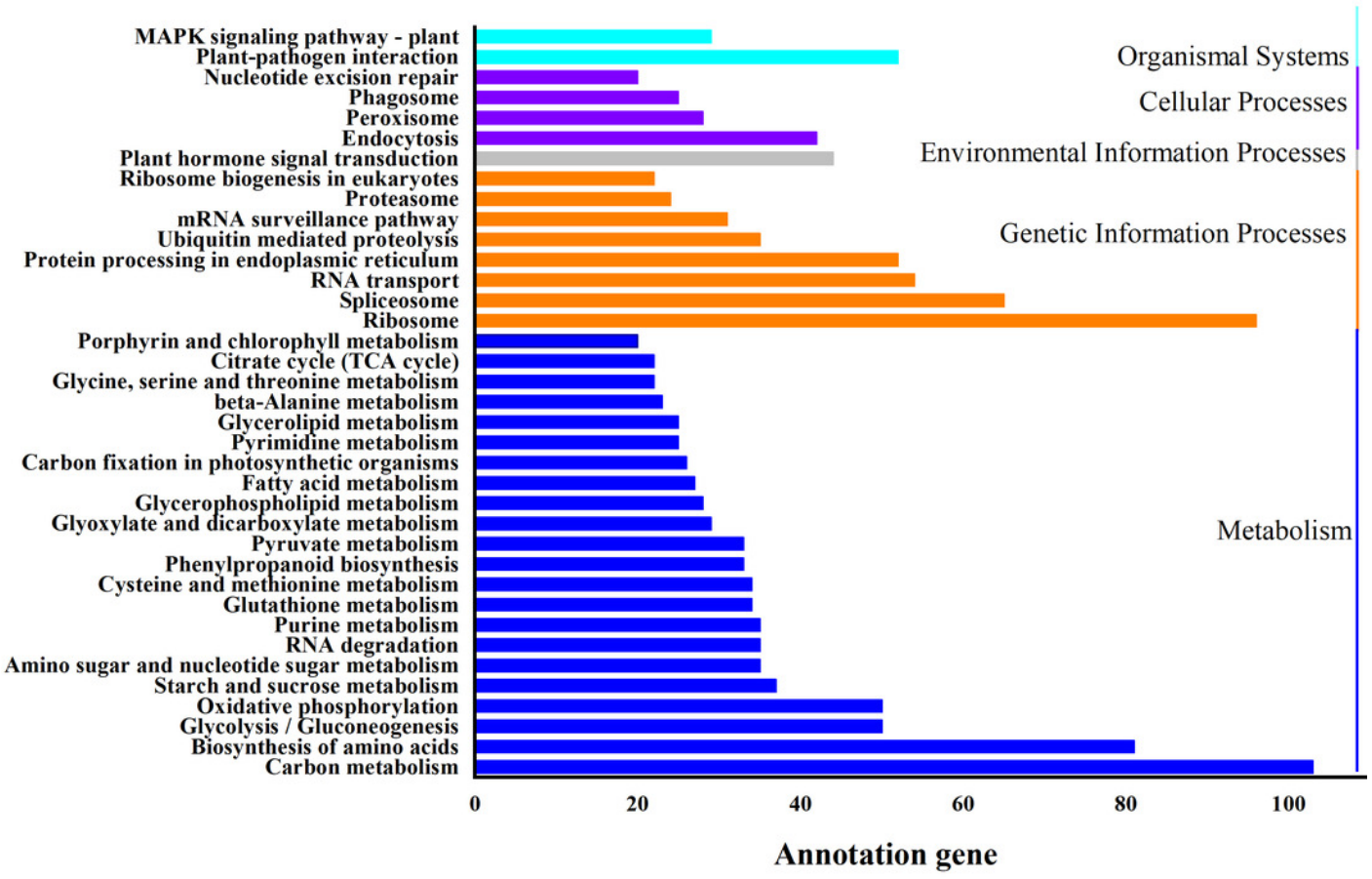

Figure 6

Bar graphs show the FPKM (fragments per kilobase of transcript per million mapped reads) of DEGs involved in ~~eight~~ biological processes distinguished by GO enrichment analysis.

~~FPKM is one of the indicators to measure gene expression level.~~ (a) ~~Bios~~ynthesis and degradation of lignin; (b) Formation of the primary cell wall or secondary cell wall; (c) Cell enlargement and plant growth; (d) Senescence.

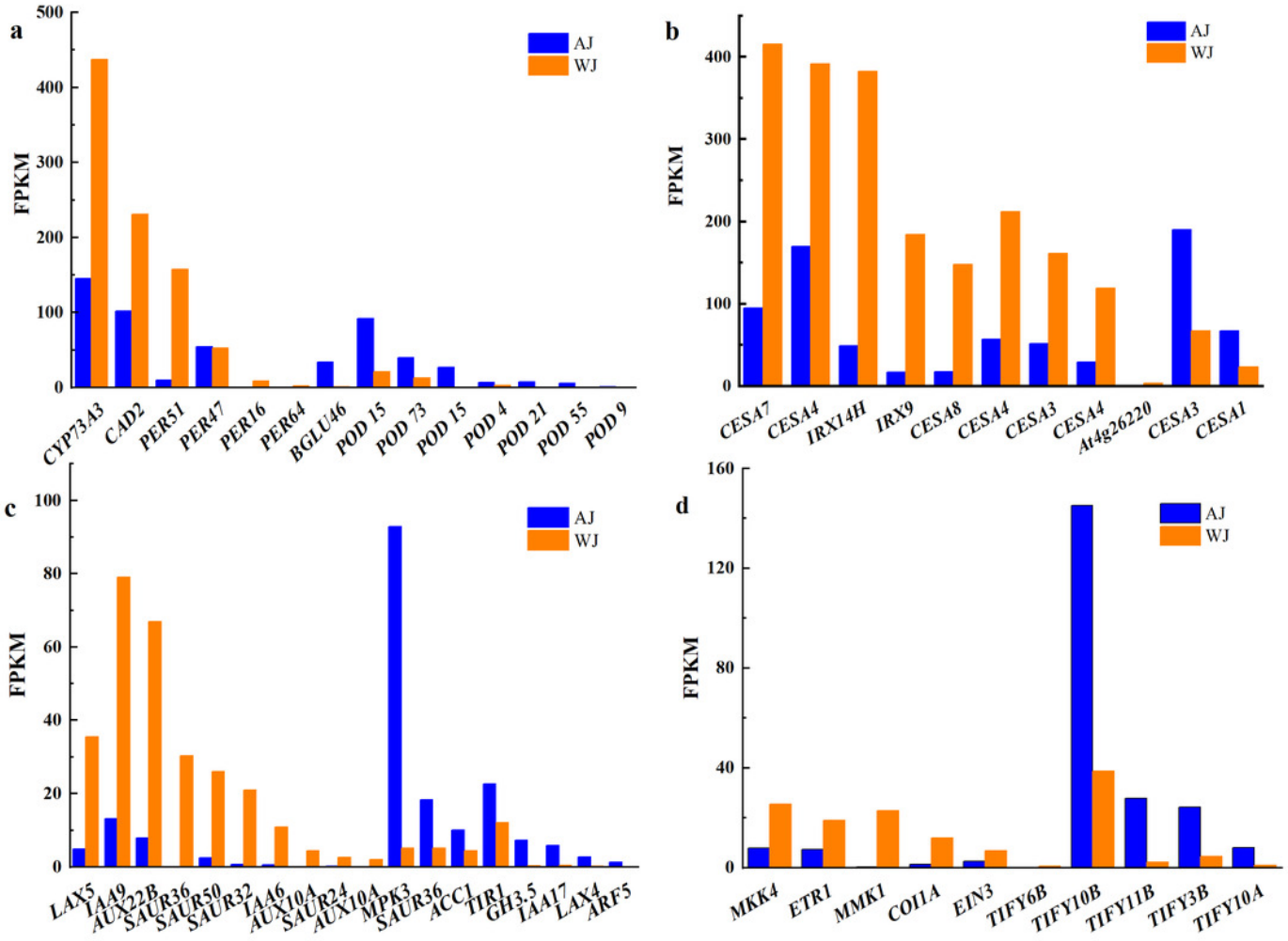

Figure 7

Bar graphs show the FPKM of DEGs involved in ~~eight~~ biological processes distinguished by GO enrichment analysis.

(a) Cell division and shoot initiation; (b) Stem growth and induced germination; (c) Cell elongation; (d) Formation of water-conducting tissues.

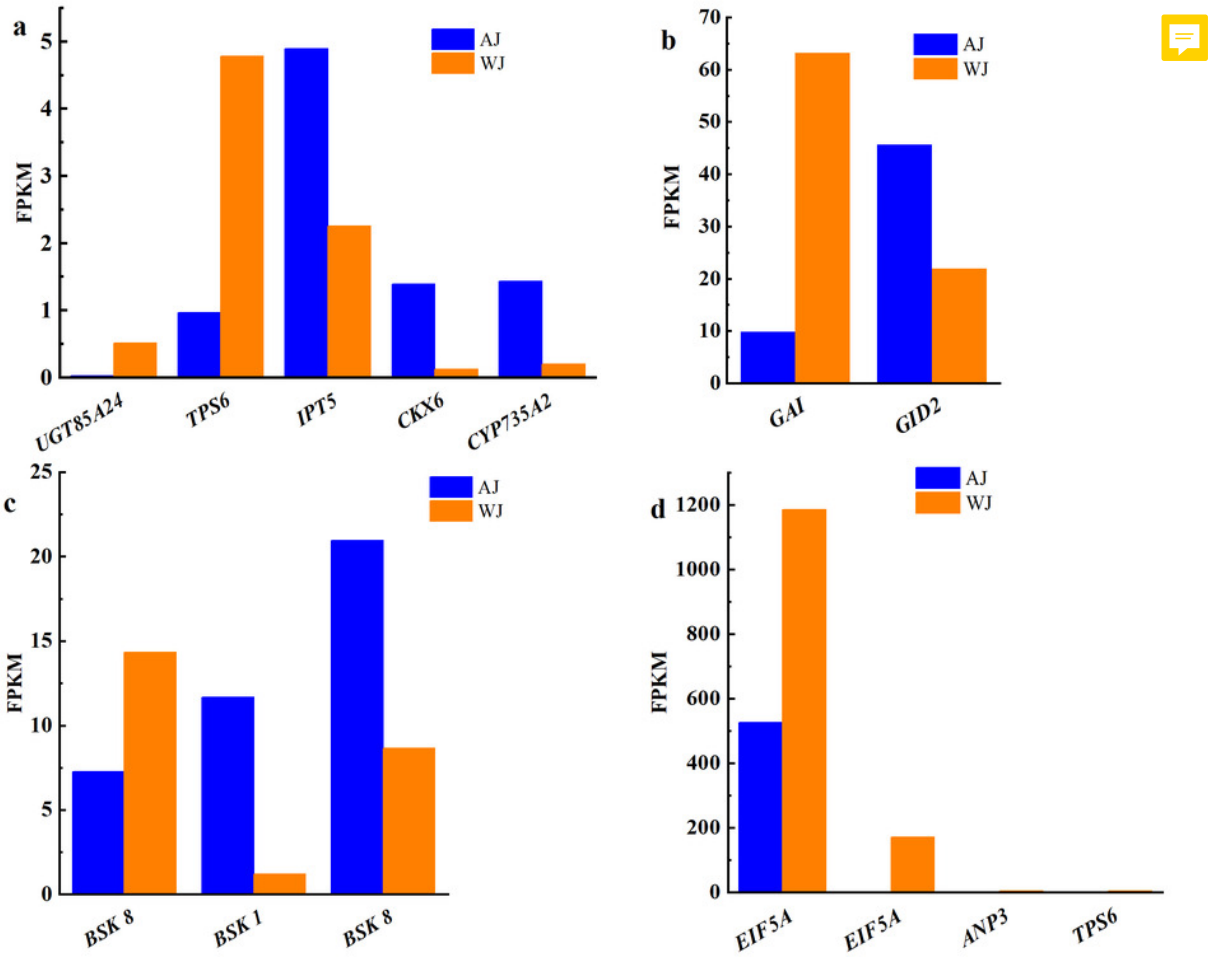

Figure 8

Transcription factors involved in stem elongation and diameter enlargement in alfalfa distinguished by GO analysis.

(a) Bar graphs show the ~~nine DEGs~~ transcript abundance ~~changes calculated from variousby~~ FPKM ~~methods.~~(b) Bars plot the ~~nine DEGs~~ relative expression levels ~~of AJ and WJ~~. *, **, *** ~~Significant~~ at the 0.05, 0.01, and 0.001 probability levels, respectively.

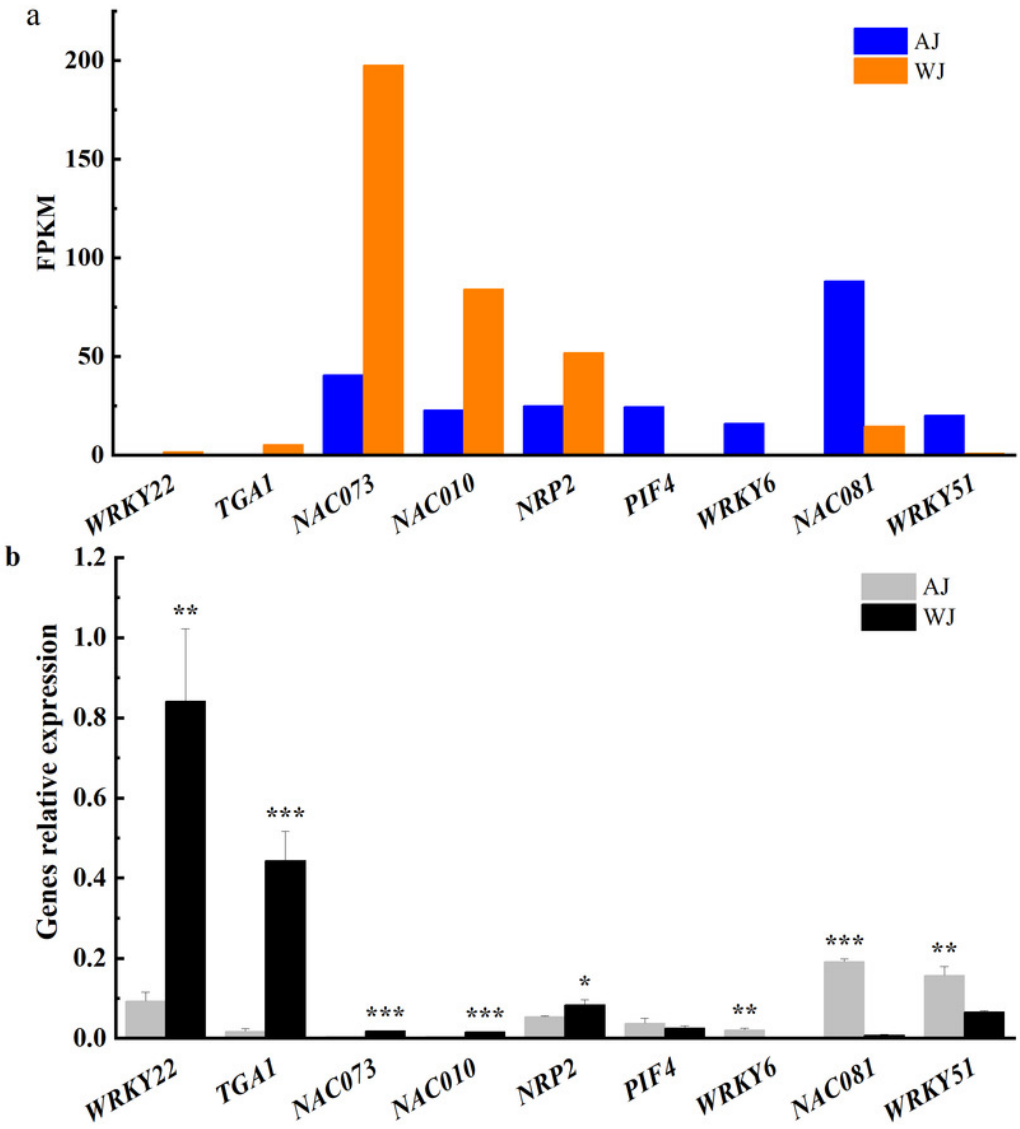

Figure 9

~~Validation by qRT-PCR of DEGs from the Aohan (AJ) and WL 712 (WJ) RNA-seq databases in alfalfa.~~

(a) Bars show ~~the eleven DEGs~~ transcript abundance ~~changes calculated from various by FPKM methods.~~ (b) Bars show ~~the eleven DEGs relative expression levels of AJ and WJ validated by qPCR. *, *** Significant~~ at the 0.05, and 0.001 probability levels, respectively.

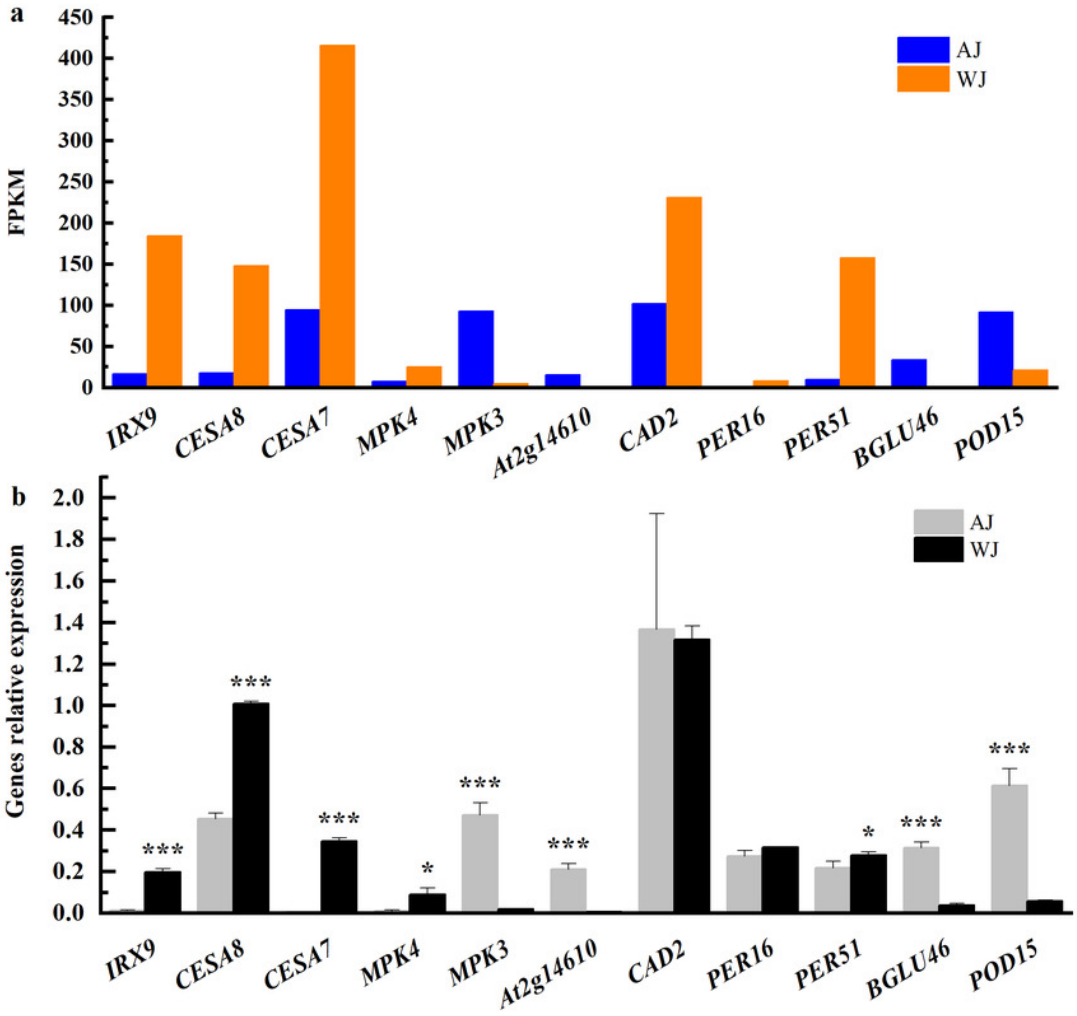

Table 1(on next page)

~~Phenotypic c~~orrelation coefficients between traits ~~based on~~ the five alfalfa cultivars

*, ** Significant at the 0.05, and 0.01 probability levels, respectively. PH, plant height; IL, internode length; SD, stem diameter; FW, fresh weight; LSR, leaf-to-stem ratio; DW, dry weight; LBN, lateral branch number; MBN, main branch number.

Table 1. Phenotypic correlation coefficients between traits based on the five alfalfa cultivars

	PH	SD	IL	LBN	MBN	FW	LSR	DW
PH	1							
SD	0.98**	1						
IL	0.99**	0.98**	1					
LBN	0.89**	0.92**	0.90**	1				
MBN	- 0.84**	- 0.76**	- 0.79**	- 0.68**	1			
FW	0.98**	0.98**	0.98**	0.91**	-0.75**	1		
LSR	0.55**	0.54**	0.55**	0.51**	- 0.46*	0.60**	1	
DW	0.99**	0.99**	0.99**	0.82**	- 0.76**	1.00**	0.59**	1

*, ** Significant at the 0.05, and 0.01 probability levels, respectively.

PH, plant height; IL, internode length; SD, stem diameter; FW, fresh weight; LSR, leaf-to-stem ratio; DW, dry weight; LBN, lateral branch number; MBN, main branch number.

Table 2(on next page)

The growth index of the two varieties in greenhouse

Different letters indicate significant difference at $P < 0.05$ among the two varieties as determined by Student's t test.

Table 2 The growth index of the two varieties in greenhouse

	Plant Height (cm)	Lenght of Internodde (cm)	Stem Diameter (mm)	Leaf Areas (mm^2)	Plant Weight (g/plant)
WL 712	50.2 ± 1 [a]	5.14 ± 0.09 [a]	2.52 ± 0.022 [a]	159 ± 0.6 [a]	231 ± 2.4 [a]
Aohan	28.7 ± 1 [c]	2.94 ± 0.07 [c]	1.19 ± 0.027 [c]	127 ± 2.8 [c]	141 ± 0.4 [c]

Different letters indicate significant difference at $P < 0.05$ among the two varieties as determined by Student's t test.

Table 3(on next page)

Top 10 gene ontology function classification

Table 3 Top 10 gene ontology function classification

Category	Description	GO ID	Count	Percentage(%)
Biological process	translational elongation	GO:0006414	41	4.30
	regulation of protein complex disassembly	GO:0043244	8	0.84
	regulation of translation	GO:0006417	7	0.73
	regulation of translational elongation	GO:0006448	7	0.73
	regulation of translational termination	GO:0006449	7	0.73
	translational frameshifting	GO:0006452	7	0.73
	posttranscriptional regulation of gene expression	GO:0010608	7	0.73
	positive regulation of cellular protein metabolic process	GO:0032270	7	0.73
	regulation of cellular amide metabolic process	GO:0034248	7	0.73
	positive regulation of cellular amide metabolic process	GO:0034250	7	0.73
Cell component	bounding membrane of organelle	GO:0098588	57	5.97
	peptidase complex	GO:1905368	44	4.61
	whole membrane	GO:0098805	49	5.13
	proton−transporting two−sector ATPase complex	GO:0033177	20	2.10
	Golgi apparatus part	GO:0044431	36	3.77
	Golgi apparatus	GO:0005794	36	3.77
	proteasome core complex	GO:0005839	37	3.88
	COPI−coated vesicle membrane	GO:0030126	9	0.94
	COPI−coated vesicle	GO:0030137	9	0.94
	COPI vesicle coat	GO:0030126	9	0.94
Molecular function	translation elongation factor activity	GO:0003746	41	4.30
	translation factor activity, RNA binding	GO:0008135	86	9.01
	acid−amino acid ligase activity	GO:0016881	12	1.26
	UDP−glycosyltransferase activity	GO:0008194	99	1.04
	threonine−type endopeptidase activity	GO:0004298	37	3.88
	threonine−type peptidase activity	GO:0070003	37	3.88
	protein heterodimerization activity	GO:0046982	114	11.95
	ligase activity, forming carbon−nitrogen bonds	GO:0016879	35	3.67
	acetylglucosaminyltransferase activity	GO:0008375	35	3.67
	oxidoreductase activity	GO:0016638	12	1.26