Transcriptome profiling of two rice genotypes under mild field drought stress during grain-filling stage

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Abstract

Drought is one of the most critical abiotic stresses that threaten crop production worldwide. This stress affects the rice crop in all stages of rice development; however, the occurrence during reproductive and grain-filling stages has the most impact on grain yield. Although many global transcriptomic studies have been performed during the reproductive stage in rice, very limited information is available for the grain-filling stage. Hence, we intend to investigate how the rice plant responds to drought stress during the grain-filling stage and how the responses change over time under field conditions. Two rice genotypes were selected for RNA-seq analysis: ‘4610’, previously reported as a moderately tolerant breeding line, and Rondo, an elite indica rice cultivar susceptible to drought conditions. Additionally, 10 agronomic traits were evaluated under normal irrigated and drought conditions. Leaf tissues were collected during grain-filling stages at two time points, 14 and 21 days after the drought treatment, from both the drought field and normal irrigated field conditions. Based on agronomic performances, ‘4610’ was less negatively affected than Rondo under mild drought conditions, and expression profiling largely aligned with the phenotypic data. The transcriptomic data indicated that, in general, ‘4610’ had much earlier responses than its counterpart in mitigating the impact of drought stress. Several key genes and gene families related to drought stress or stress-related conditions were found differentially expressed in this study, including transcription factors, drought tolerance genes and reactive oxygen species scavengers. Furthermore, this study identified novel differentially expressed genes (DEGs) without function annotations that may play roles in drought tolerance-related functions. Some of the important DEGs detected in this study can be targeted for future research.

Keywords: Differentially expressed genes (DEGs); drought; grain-filling stage; rice (Oryza sativa); RNA-seq.

Introduction

Rice (Oryza sativa L.) is a staple food that feeds about 3.5 billion people, more than half of the world’s population (Ricepedia 2013). The global population is projected to reach 9.1 billion by 2050 (Food and Agriculture Organization of the United Nations 2009); therefore, an increase in global rice production is critical to keep pace with the rising global demand for food. However, rice production faces daunting challenges, including abiotic stress problems. Drought is one of the main obstacles to rice production; nearly 50% of rice farming is affected worldwide (Bouman et al. 2005). This stress can cause limited nutrient uptake, cell dehydration and the production of excessive metabolic compounds of reactive oxygen species (ROS).
(Price et al. 1989; Choudhury et al. 2017). Simple cell dehydration may damage cells, including membrane dysfunction, misfolded protein and cytoskeletal damage. Likewise, ROS may cause DNA, protein and membrane damages (Mittler 2002; Tenhaken 2015).

Drought tolerance mechanisms are complex. In legumes, although drought ultimately affects the productivity of grains at all growth stages, the incidence during reproductive and grain development stages results in more substantial grain production loss (Farooq et al. 2017). A similar phenomenon was also observed in cereals, such as wheat and rice (Tripathy et al. 2000; Farooq et al. 2014; Prathap et al. 2019). Drought stress during the reproductive stage can reduce yield dramatically (Garrity and O’Toole 1994; Venuprasad et al. 2007; Raman et al. 2012). Likewise, many studies have shown that drought stress also has significant impacts during the grain-filling stage, including decreased photosystem II (PSII) activity (Pieters and El Souki 2005), rapid accumulation of free radical (O$_2^•^−$) and hydrogen peroxide (H$_2$O$_2$) (Wang et al. 2010), fewer sink activities and smaller sink sizes (Dong et al. 2014) and early senescence leading to decreased starch in matured grains and the loose packaging of the starch granules (Prathap et al. 2019).

There are two main categories of drought tolerance mechanisms, preventing water loss and preventing damage caused by dehydration (McDowell et al. 2008; Farooq et al. 2009). Abscisic acid (ABA) regulation (Munemasa et al. 2015), K$^+$ concentration and membrane potential (Corraige-Faille et al. 2017) can regulate stomatal conductance to reduce water loss (Chaves et al. 2008). Overexpression of drought-induced genes such as calcium-dependent protein kinase (CDPK), NAM, ATAF and CUC (NAC) transcription factors (TFs), and dehydration-responsive element-binding proteins (DREBs) contribute to the drought tolerance response in rice (Saijo et al. 2000; Hu et al. 2006; Chen et al. 2008).

Rice transcriptomic studies under drought conditions in different scenarios have been reported, including during seedling stage and reproductive stage (Lenka et al. 2011; Huang et al. 2014; Plessis et al. 2015; Wilkins et al. 2016; Zhang et al. 2016; Yoo et al. 2017). To our knowledge, however, transcriptome profiling of drought response mechanisms during grain-filling stage is very limited. Hence, our current study aimed to investigate the molecular mechanisms underlying drought tolerance during the grain-filling stage and to evaluate how the responses change during this period under realistic field conditions. A time series of RNA-seq using two rice genotypes having different responses to drought, i.e. Rondo and ‘4610’, was performed. Additionally, some agronomic traits were measured. Our results showed that, in general, earlier cellular events, signalling cascades and stress-responsive mechanisms in ‘4610’ than Rondo were observed. Overall, both transcriptome profiling and phenotypic data concurred that ‘4610’ was more tolerant under drought stress. Some of the differentially expressed genes (DEGs) detected in this study can be further investigated for further molecular studies and manipulation for crop improvement to enhance plant survival and mitigate grain yield reduction under drought conditions.

Materials and Methods

Plant materials and drought treatment

Two rice genotypes with different performances under drought stress, Rondo and ‘4610’, were selected for this study. Rondo is an elite indica long-grain rice cultivar, developed by the USDA-ARS through mutation breeding from the Chinese indica germplasm ‘4484’ (PI 615022). This cultivar has a high yield, is resistant to various diseases (Yan and McClung 2010) and is mainly grown in the mid-south rice belt of the USA, including Arkansas, Louisiana, Mississippi, Missouri and Texas. Rondo is also being planted for commercial organic rice production in Texas. However, this cultivar is susceptible to drought stress (Tabien et al. 2015; Dou et al. 2016). On the other hand, ‘4610’ is an indica long-grain (PI615037) introduced from China (Dilday et al. 2001), previously selected from herbicide-resistant screening at the Texas A&M AgriLife Research Center in Beaumont, TX, USA, and was reported as less affected by drought (Tabien et al. 2015).

It is intriguing to understand how these two different cultivars respond to field drought stress and how to use this information to enhance the tolerance of the elite cultivar Rondo and possibly other susceptible indica cultivars or breeding lines in the future.

A field experiment was performed at the Texas A&M AgriLife Beaumont Research Center in 2016 and 2017. Beaumont is located in East Texas and has an average of 18–31.6 °C daytime temperature and 13.72 cm precipitation per month during June–October, the duration of this experiment. Unfortunately, our experiment in 2017 was affected by the tropical storm Harvey. Considering that Rondo and 4610 had previously been screened under drought conditions at the same experimental station (Tabien et al. 2015; Dou et al. 2016), we envisage that our data can be used as further phenotypic confirmation of previous screening results.

The two rice genotypes were directly sowed in the experimental rice field using a randomized complete block design (RCBD) with three replications in normal irrigated and drought conditions. The rates and timing of fertilizer and herbicide were similar for both fields. A total of 56 kg P per hectare (as P$_2$O$_5$) was applied during pre-season, and 224 kg ha$^{-1}$ N was applied as urea as follows: 56 pre-plant, 90 at the permanent flood and 78 at panicle differentiation. Glyphosate was applied pre-season; clomazone + halosulfuron-methyl post-plant, and cyhalofop at the permanent flood and 14 days later. The heavy clay soil was representative of rice farms in the area. All replicates were planted in three-row blocks; each block has 6 m in length and 18 cm between rows. Drought treatment was applied once >50% of plants had flowered by draining the field completely. There was a light rain a few days after the draining; however, we were able to manage and maintain the drought conditions afterward and performed the tissue sampling for RNA-seq analysis on the 14th day and 21st day after the draining. The drought field was then rewatered right after the second tissue sampling.

Phenotypic evaluation

A total of 10 agronomic traits were measured at physiological maturity; these were filled grain number per panicle, unfilled grain number per panicle, spikelet fertility, filled grain weight per plot, unfilled grain weight per plot, total grain weight per plot, plant dry weight, hundred-seed weight, panicle length and yield. The 10 representative panicles from each plot were used to count the grain number, including filled and unfilled grain numbers. Spikelet fertility was calculated from filled and unfilled grain numbers on each of the selected panicles. All panicles in each plot were manually threshed and manually separated as filled and unfilled grains. Filled and unfilled grain data were then used to calculate filled grain weight, unfilled grain weight and total grain weight per plot. Plant dry weight was measured for the above-ground part of the plant after air-drying for 7 days according to the protocol of the Beaumont Station Rice Breeding Lab. Hundred-seed weight was the average of three subsamples of 100 randomly selected filled grains. Panicle length was measured as average from 20 panicles in each plot. The yield was counted from the total grain weight.
The regions and obtain the gene counts.

RNA extraction and sequencing

RNA was extracted from leaf tissues collected at two different time points, 14 and 21 days after the field for the drought treatment was drained. These time points corresponded to the earlier and later grain-filling stages, respectively. Early symptoms of drought (leaf rolling) were observed during the samplings at both time points. It was also observed that there were no drastic differences in the severity of drought across both time points; the level of drought stress at both time points was mild. Leaf tissue samples were also taken from the normal irrigated field at the same time points. The samplings were performed in the morning, around 10–11:30 am. For each condition and each time point, three biological replicates were collected. These replicates were from three different plants, and the leaves collected on the 14th day and the 21st day were from the same plant. The second leaves, counted from the top, on the main stem were collected on the 14th day. The third leaves, counted from the top, on the main stem were collected on the 21st day. RNA was extracted using TRIzol reagent, followed by QiAgen RNaseasy Plant Mini Kit. The samples were not DNase-treated; however, based on the assessment on agarose gel, there was no DNA contamination in any of the RNA samples. RNA quality was determined by 28S to 18S ribosomal RNA ratio and samples with a ratio within the range of 1.8–2.2 were checked further using Agilent 2100 Bioanalyzer (Santa Clara, CA, USA). The samples that passed the quality check were used for the library preparation. TrueSeq Stranded RNA-seq libraries were prepped at Texas A&M AgriLife Genomics and Bioinformatics Service (TxGen; College Station, TX, USA) as per Standard Operating Protocol (Illumina). The libraries were run on multiple lanes of an Illumina HiSeq 4000 (San Diego, CA, USA) to provide at least 25 million reads (75 nt pair-end) per sample.

Data processing

In total, 12 CDNA libraries were made for sequencing, including three biological replicates for each treatment. Trimmmomatic version 0.36 (Bolger et al. 2014) PE –Phred 33 command was used for quality control using the following steps: (i) raw sequencing reads were trimmed to remove adaptors; (ii) low-quality bases with a quality score less than 20 on the ends and tails of reads were removed; (iii) reads were scanned with a 5-bp sliding window and were removed when the average quality per bp dropped below 20; (iv) reads below the 25 bases long were dropped; (v) reads without correspondence read pairs were dropped. Oryza sativa spp. japonica genome Nipponbare (International Rice Genome Sequencing Project (IRGSP)-1.0; Sasaki and Burr 2000; Kawahara et al. 2013) was used as a reference in this study. The rice reference sequence and gene annotation files were downloaded from EnsemblPlants (http://plants.ensembl.org). HISAT2 version 2.1.0 (Kim et al. 2015) was used to align the reads to the reference genome sequence. Thereafter, StringTie v1.3.4d (Pertea et al. 2015) was used to assemble the transcripts within the regions and obtain the gene counts.

Differential expression and gene ontology enrichment analyses

Differential expression analysis was performed in R studio using DESeq2 v1.26 (Love et al. 2014) with reads normalizing and variance stabilizing transformation to account for library size and sequencing depth differences. A generalized linear model of \( Y = \gamma_{\text{geno}} + \gamma_{\text{trt}} + \gamma_{\text{geno} \times \text{trt}} \), where \( Y \) is the read count for each gene, with three explanatory variables, including genotype, treatment and interaction between genotype and treatment, was used in this study. Due to the different growing stages, data from the two time points were fitted in the model separately. Internally, \( P \)-values were adjusted for multiple testing using the Benjamini–Hochberg method in the DESeq2 package. Genes with false discovery rate (FDR) adjusted \( P \)-value \( (p_{\text{adj}}) < 0.05 \) were identified as DEGs. There were six DEG lists used for the gene ontology (GO) enrichment analysis in both reproductive and grain-filling stages, including uniquely upregulated and downregulated DEGs in ‘4610’, uniquely upregulated and downregulated DEGs in Rondo, commonly upregulated DEGs in both cultivars and commonly downregulated DEGs in both cultivars. Gene ontology enrichment analysis was performed using the web-based tool PANTHER v15.0 (Mi et al. 2019). For GO enrichment analysis, Fisher’s Exact test was used for the main statistical analysis with FDR adjusted by the Benjamini–Hochberg method. Gene ontology terms with FDR < 0.05 were identified as significant.

Results and Discussion

Agronomic performance under mild field drought stress

Agronomic characters were collected from 12 plots, covering three replicates for each genotype under each condition. Under the irrigated condition, Rondo had a significantly higher number of both filled and unfilled grains than those of ‘4610’. The panicle traits with ‘*’ and ‘**’ indicates that the difference between the two cultivars under the same condition was significant by \( P < 0.05 \) and \( P < 0.001 \), respectively.

| Table 1. Effect of variety on yield-related traits under irrigated and drought condition. |
|---------------------------------|-------------------|-------------------|
|                                | Irrigated ‘4610’  | Drought ‘4610’   |
| (1) Filled grain number per panicle | 68.2*             | 69.5*             |
| (2) Unfilled grain number per panicle | 21.2*             | 31.2              |
| (3) Spikelet fertility (%)        | 76.9              | 69.5*             |
| (4) Filled grain weight (g) per plot | 77.3              | 55.3              |
| (5) Unfilled grain weight (g) per plot | 5.7               | 4.1               |
| (6) Total grain weight (g) per plot | 83.0              | 59.5              |
| (7) Plant dry weight (g)          | 200.76            | 126.6             |
| (8) Hundred-seed weight (g)       | 2.9**             | 2.9**             |
| (9) Panicle length (cm)           | 19.5**            | 19.5              |
| (10) Yield (kg ha\(^{-1}\))      | 10,666            | 9,850.04          |
Principal component analysis of RNA-seq samples and the number of DEGs

There were three explanatory variables in the current study, and these were genotype (‘4610’ and Rondo), treatment (irrigated and drought) and time point (14 days and 21 days after drought treatment). To explore the comprehensive gene expression differences of the drought susceptible elite cultivar Rondo and the moderate-tolerant breeding line ‘4610’ under different factors, a total of 38,909 annotated rice genes with at least 10 counts were used for principal component analysis (PCA) as the first step of data exploration. As sources of variations were explored, the different sample collection time points appeared to be the most significant factor in explaining gene expression variations. For the first principal component (PC), which explained 40% of the total variance (Fig. 1), the sample collection time points showed the dominant loading. The second most important factor appeared to be genotype, which displayed the dominant loading for PC2, which explained 26% of the total variance. As samples from the same genotype grouped close to each other regardless of the irrigated or drought condition, meaning samples with the same genotype had relatively similar expression patterns.

To minimize the effect of different time points, data from 14- and 21-day time points were analysed independently. Principal component analysis separated by the two time points suggested that genotype explained 40% of the total variance on Day 14 and 38% of the total variance on Day 21 [see Supporting Information—Fig. S1]. A Venn diagram illustrated the number of DEGs for the four different comparisons (Fig. 2). A total of 1107 and 925 genes were identified to be differentially expressed under drought stress for ‘4610’ and Rondo at 14-day time point with 552 and 490 upregulated genes, and 555 and 435 downregulated genes in ‘4610’ and Rondo, respectively [see Supporting Information—Table S1]. Among all DEGs, there were 70 upregulated and 42 downregulated DEGs in common (Fig. 2). For the 21-day time point, there were 679 and 5099 DEGs identified, with 278 and 2635 upregulated genes and 555 and 435 downregulated genes in ‘4610’ and Rondo, respectively. Among all DEGs, there were 167 upregulated and 227 downregulated DEGs commonly identified in both genotypes (Fig. 2).

It was interesting to note that, in general, 4610 had a higher number of DEGs at the earlier stage of grain filling than Rondo; on the other hand, Rondo had a much higher number of DEGs than 4610 on the later stage. This phenomenon suggests that 4610 may give a more rapid response to drought stress than Rondo. However, at the later stage, it seems that Rondo gave an ‘overacting’ response to the stress, which in turn may negatively impact the plant’s cellular homeostasis. This occurrence may also partially explain a better performance in 4610 under drought stress compared to Rondo. A more rapid and stronger drought-responsive regulation in drought-resistant upland rice than a lowland drought susceptible cultivar was also reported in a previous study (Teng et al. 2014). A study in barley revealed that a rapid upregulation of dehydrin under dehydration conditions is the a key characteristic of drought-tolerant cultivar (Park et al. 2006). Our analysis also showed that at earlier time point dehydrin rab (responsive to ABA) 16C (Os11g0454000) and Rab21 (Os11g454300) were strongly
upregulated in 4610, but not in Rondo (Fig. 3). Here, we were interested in further dissecting the underlying molecular genetic factors that may contribute to the performance of the cultivars under drought conditions.

Molecular response to drought at the early stage of grain filling (14-day time point)

There were 106 DEGs identified among genotype and treatment at 14-day time point [see Supporting Information—Tables S1 and S3], including some TFs (Os12g0155500, Os01g0584900, Os05g0322900, Os04g0656500 and Os09g0434500), stress-related genes (the stress upregulated Nod 19 family protein (Os08g0538600), a gene related to calcium signalling during abiotic stress (Os01g0135700), and a gene contributing to the regulation of abiotic stress responses (Os07g0129200) and transporters (Os02g0518600, Os03g0226400 and Os09g0434500)). To understand the functional classification underlying these unique DEGs, GO enrichment analysis was performed. There were 12 GO terms enriched in biological process (BP), including response to stimulus (GO:0009611), response to stress (GO:0006950), defence response (GO:0006952) and response to lipid, and 10 GO terms enriched in molecular function (MF), including hormone binding (GO:0042562), asbicsic acid binding (GO:0010427) and isoprenoid binding (GO:0019840). However, there were no cellular component (CC) GO terms enriched [see Supporting Information—Table S3].

Among 482 uniquely upregulated DEGs in ‘4610’, there are 20 TFs and 17 stress-related genes. Gene ontology analysis further revealed that a total of 76 GO terms were enriched, including 58 BP, 13 MF and 5 CC [see Supporting Information—Table S3]. Among 76 GO terms, 14 of them were stress-related (Table 2). On the contrary, none of these GO terms were significant in upregulated DEGs in Rondo. Among 420 uniquely upregulated DEGs in Rondo, there were 73 GO terms enriched, including 42 BP, 14 MF and 17 CC [see Supporting Information—Table S3]. Among 70 commonly upregulated DEGs including one bZIP TF (Os01g0658900), five known drought-responsive genes including late embryogenesis abundant (LEA) proteins (Os01g0705200, Os11g0454200), RIC2 family protein (Os03g0286900), ascorbate peroxidase (APX) (Os04g0434800) and drought and salt stress response 1 (Os09g0109600), and two heat shock proteins (HSP) (Os02g0232000, Os03g0277300) [see Supporting Information—Table S1]. There were only six GO terms enriched in CC, including those that play roles in the vacuole and membrane-related functions, such as vacuolar iron transporter (Os04g0538400), mitochondrial import inner membrane translocase subunit (Os03g0305600) and DDHD domain-containing protein (Os08g0110700). Several studies have shown that vacuole and vacuole transporters involve in complex cellular networks and facilitate plants to adjust environmental changes, including salinity tolerance, calcium signalling and cellular pH homeostasis (Martinoia et al. 2007). On the other hand, ROS can be generated at the plasma membrane, which may affect membrane stability (Ijmpa et al. 2012). In addition, H2O2 can also cross plant membranes and further activate downstream stress-responsive signalling (Pitzschke et al. 2006). However, there was no enrichment on GO terms for BP and MF [see Supporting Information—Table S3].

Downregulated DEGs were also enriched in distinct categories between ‘4610’ and Rondo. There were 101 BP, 37 MF and 45 CC enriched among 513 uniquely downregulated DEGs in ‘4610’, whereas Rondo had 61 BP, 34 MF and 8 CC among 393 unique DEGs [see Supporting Information—Table S3]. Interestingly, several photosynthesis-related GO terms were enriched in ‘4610’, including PSI repair (GO:0010206), chlorophyll biosynthetic process (GO:0015995), photosynthesis, light reaction (GO:0019684), photosynthesis (GO:0015979) and chlorophyll metabolic process (GO:0015994). Conversely, most of these photosynthesis-related GO terms were not enriched in Rondo downregulated DEGs, whereas the response to stress (GO:0006950) was enriched with 43 DEGs [see Supporting Information—Table S3]. Different mechanisms exist through which the rate of photosynthesis under drought stress is suppressed. For example, stomatal closure, diminished CO2 influx, decreased Rubisco activity and ROS accumulation (Farooq et al. 2009). A strong positive selection on photosynthetic genes conferred upland rice with better drought tolerance than lowland rice, with less photosynthesis rate decline (Zhang et al. 2016). The different results might be due to various factors such as plant growth stage and genotype. Many drought transcriptomics studies in rice were conducted during the seedling, vegetative or reproductive stages (Chen et al. 2017; Lenka et al. 2011; Huang et al. 2014; Zhang et al. 2016; Borah et al. 2017; Tarun et al. 2020); however, the responsive mechanisms may differ during the grain-filling stage and need further investigation. A previous study showed that both tolerant and susceptible cultivars had severely decreased photosynthesis rates. The increased abundances of ROS scavengers such as chloroplastic superoxide dismutase and dehydroascorbate reductase, which provide antioxidant protection against damage by dehydration, contributed to tolerance (Ji et al. 2012). The two varieties may also have unique adaptation or tolerance mechanisms in response to drought stress. Only 42 DEGs were commonly downregulated in both cultivars with 9 BP and 2 MF terms were significantly enriched. Among the common DEGs, four of them were directly related to yield-component traits such as panicle-branching, grain weight and grain size [see Supporting Information—Tables S1 and S3].

Molecular response to drought during the later stage of grain filling (21-day time point)

There were 280 DEGs identified having interaction between genotype and treatment, including some TFs, stress-responsive genes and HSPs at 21-day time point [see...
Figure 3. The expression differences of known drought-related DEGs between ‘4610’ and Rondo, in reproductive (Day 14) and grain-filling stages (Day 21), including TFs, cytochrome P450 families, LEA genes, ROS scavengers and other drought-related genes. Heatmap labelled in white indicates log2 fold-change (LFC) equal to 0, meaning the expression level had no difference between control and drought conditions. A higher LFC means the gene had a higher expression level compared to the same gene under control condition.
### Table 2. Stress-related GO categories of ‘4610’ uniquely upregulated DEGs during reproductive stage.

| Biological process                                      | GO ID      | Number of genes in ‘4610’ | FDR           | Number of genes in Rondo | FDR  |
|---------------------------------------------------------|------------|---------------------------|---------------|--------------------------|------|
| Response to temperature stimulus                        | GO:0009266 | 21                        | 5.56E-15      | 5                        | ns   |
| Response to abiotic stimulus                            | GO:0009628 | 29                        | 5.71E-09      | 12                       | ns   |
| Response to heat                                        | GO:0009408 | 13                        | 9.69E-08      | 3                        | ns   |
| Response to osmotic stress                              | GO:0006970 | 14                        | 3.13E-07      | 1                        | ns   |
| Response to salt stress                                 | GO:0009651 | 12                        | 4.15E-06      | 1                        | ns   |
| Oxidation-reduction process                             | GO:0055114 | 50                        | 2.07E-05      | 37                       | ns   |
| Response to stimulus                                    | GO:0050896 | 63                        | 2.32E-04      | 57                       | ns   |
| Response to ROS                                         | GO:0003030 | 8                         | 2.33E-04      | 2                        | ns   |
| Response to stress                                      | GO:0006950 | 36                        | 3.90E-03      | 23                       | ns   |
| Response to water deprivation                           | GO:0009414 | 7                         | 4.03E-03      | 3                        | ns   |
| Positive regulation of transcription from RNA polymerase II promoter in response to stress | GO:0036003 | 4                          | 2.28E-02      | 2                        | ns   |
| Glutathione metabolic process                           | GO:0006749 | 6                         | 3.15E-02      | 0                        | ns   |
| Regulation of DNA-templated transcription in response to stress | GO:0043620 | 4                          | 3.49E-02      | 2                        | ns   |

Supporting Information—Table S2]. These DEGs were enriched in 27 BP, 47 MF and 17 CC GO terms [see Supporting Information—Table S4]. The BP terms included a response to H₂O₂ (GO:0042542), response to ROS (GO:0000302), chaperone-mediated protein folding (GO:0006107) and protein folding (GO:0006457), and the MF included several membrane activity-related GO terms [see Supporting Information—Table S4]. Notably, there were chloroplast stroma (GO:0009570), plastid stroma (GO:0009532) and chloroplast (GO:0009507) among CC terms [see Supporting Information—Table S4]. These indicated ‘4610’ and Rondo had different responses in various functions under drought.

There were 111 upregulated DEGs solely in ‘4610’, including a WRKYTF (Os01g0665750), two stress-related genes (Os01g0667200, Os06g0662900) and five HSPs [see Supporting Information—Table S2]. These DEGs can be functionally characterized into 15 BP, 4 MF and 1 CC terms [see Supporting Information—Table S4]. The top five BP terms with the largest fold-change were glyoxylate cycle (GO:0006097), glyoxylate metabolic process (GO:0006487), response to heat (GO:0009408), protein folding (GO:0006457) and small molecule catabolic process (GO:0044282) [see Supporting Information—Table S4]. On the other hand, 2468 uniquely upregulated DEGs in Rondo were enriched in 203 BP, 86 MF and 54 CC terms [see Supporting Information—Table S2 and S4]. The top five BP with the largest fold-change were sesquiterpene biosynthetic process (GO:0051762), sesquiterpene metabolic process (GO:0051761), ceramide biosynthetic process (GO:0046513), terpene biosynthetic process (GO:0046246) and acylglycerol metabolic process (GO:0006639). Additionally, three stress-responsive terms were also enriched, including regulation of response to stress (GO:0008134), response to salt stress (GO:0009651), response to osmotic stress (GO:0006970), response to stress (GO:0009650), response to water deprivation (GO:0009414) and response to water (GO:0009415) [see Supporting Information—Table S3].

Drought-responsive DEGs throughout the two time points of the grain-filling stage

In ‘4610’, there were 925 out of 1107 DEGs that uniquely identified in the early stage of grain filling (14-day time point) compared to the later stage (21-day time point). There were 497 out of 679 DEGs that were uniquely identified in the later stage compared to the earlier one [see Supporting Information—Table S5; Supporting Information—Fig. S2]. On the other hand, Rondo had 707 out of 925 DEGs that uniquely identified in the earlier stage of grain filling compared to the later stage of grain filling; and there were 4881 out of 5099 DEGs that uniquely identified in the later grain-filling stage compared to the earlier stage [see Supporting Information—Table S5; Supporting Information—Fig. S2].
We further examined DEGs that were common in both time points. For ‘4610’, there were 104 DEGs that were in common, while Rondo had 77 DEGs that commonly downregulated in both time points. Among 104 DEGs, 14 of them had been annotated with drought tolerance or stress response, including bZIP TF, LEA proteins, auxin-responsive protein, RIC2 family protein, GST, Rab protein, NAC-domain proteins and HSPs (see Supporting Information—Table SS).

On the other hand, Rondo had 107 upregulated DEGs and 47 downregulated DEGs consistent in both time points, with 12 DEGs were annotated as drought tolerance or stress response [see Supporting Information—Table SS]. Out of 12 stress-related DEGs, 3 DEGs were cytochrome P450 family proteins (Os02g01855200, Os02g05039000, Os03g07603000). Interestingly, none of the cytochrome P450 family proteins were identified in ‘4610’ in the earlier stage, but there were 6 DEGs of cytochrome P450 family proteins identified in the later stage [see Supporting Information—Table SS].

A total of 18 DEGs were consistently upregulated in both time points in both genotypes, with log2 fold-change from 0.58 to 3.89 in earlier stage, and log2 fold-change from 0.5 to 2.97 in later stage. There were 11 DEGs that have been annotated as stress-related DEGs, one TF, three unknown function DEGs and the other three stress-unrelated DEGs (Table 3).

Expression patterns of genes involved in signalling cascades and transcriptional control

Several known drought tolerance genes were identified in our study; some were commonly upregulated in both DEGs and time points, including OsDDR1 (Os09g01096000), Os02g08245000 (Remorin), Os01g03101000 (Phospholipase D P1, PLDP1) and the well-known drought-tolerant gene rare cold-inducible 2 (RCI2) (Li et al. 2014) (Fig. 3). OsDDR1, which stands for drought and salt stress response 1, is a small peptide protein that can be induced by several environmental stresses, including drought, salinity, ABA and H2O2 treatment, and has been shown to enhance drought tolerance by overexpression (Cui et al. 2018). Os01g03101000, annotated as OsPLDzeta2, is one of the Phospholipase D (PLD) gene family involved in lipid-mediated signalling. PLD genes can be grouped into several different categories, including alpha, beta, gamma, delta and zeta, according to their structure, and different PLD gene categories also have different functions (Qin and Wang 2002). For example, overexpression of PLDalpha was reported to enhance tolerance to drought and osmotic stress in Arabidopsis (Wang et al. 2014), PLDBeta identified as a disease resistant-related gene in rice (Yamaguchi et al. 2009) and PLDzeta1 and PLDzeta2 had distinct effects on salinity stress response in Arabidopsis (Ben Othman et al. 2017). OsPLDzeta2 was reported inducible under drought and cold stress in rice (Singh et al. 2012). OsPLDzeta2 has not yet been functionally annotated; however, based on the function of other PLD groups and the expression profile in the current study (i.e. Rondo had a higher expression level than ‘4610’), we suggest that OsPLDzeta2 may play a negative role in drought tolerance in rice (Distéfano et al. 2015; Ben Othman et al. 2017). In this study, Os02g08245000, a remorin (REM) similar gene, was first identified in rice under drought stress. It was shown that remorin group 4 (AtREM4) genes interact with SnRK1 in Arabidopsis, which is known to confer stress tolerance through sugar and hormonal signalling (Cho et al. 2012; Lastdrager et al. 2014; Son et al. 2014). Previous studies had shown that several abiotic stresses, including drought tolerance in tomato and flooding tolerance

| Gene ID          | Annotation                                          | Day 14     | Day 21     |
|------------------|-----------------------------------------------------|------------|------------|
| Os01g02003000    | Similar to Homebox-leucine zipper protein HOX29, Homeodomain transcription factor | 2.57       | 1.21       |
| Os01g03101000*   | Similar to PLDP1 (phospholipase D)                  | 0.80       | 1.18       |
| Os01g06589000*   | OSBZ8, bZIP transcription factor                    | 1.32       | 2.51       |
| Os01g07052000    | Late embryogenesis abundant (LEA) protein repeat-containing protein | 3.89       | 3.89       |
| Os02g04634001    | Conserved hypothetical protein                      | 2.04       | 0.72       |
| Os02g08245000*   | Similar to Remorin                                  | 2.87       | 1.51       |
| Os03g02773000    | Heat shock protein 70                               | 1.65       | 1.52       |
| Os03g02869000*   | RCI2 (rare cold-inducible 2) family protein, Drought resistance | 2.65       | 1.57       |
| Os03g03056000*   | Mitochondrial import inner membrane translocase, subunit Tim17/22 family protein | 2.86       | 1.55       |
| Os03g07234000    | Similar to UFG2, endosperm-specific gene 53        | 3.35       | 2.20       |
| Os05g03739000    | Similar to Eukaryotic peptide chain release factor subunit 1 (eRF1) | 1.52       | 1.50       |
| Os07g01908000*   | Similar to Thioredoxin h                            | 1.01       | 1.00       |
| Os07g06400000*   | Similar to 6-phosphogluconolactonase-like protein   | 1.38       | 0.86       |
| Os08g04033000    | Heavy metal transport/detoxification protein domain-containing protein | 1.71       | 2.77       |
| Os08g04258000    | Conserved protein, expressed                        | 0.71       | 0.58       |
| Os09g01096000*   | Small peptide, Drought tolerance                    | 2.76       | 2.17       |
| Os09g05724000*   | ABC transporter, ATP-binding protein, putative, expressed | 0.60       | 0.79       |
| Os11g05334000    | Conserved hypothetical protein                      | 3.07       | 1.27       |
during germination in rice, could be enhanced by increasing interaction with SnRK1 (Kretzschmar et al. 2015; Yu et al. 2018; Yu et al. 2021). According to the expression profile, ‘4610’ had relatively higher expression than Rondo in both time points (Fig. 3). This suggests that Os2g0824500 may also play a role in drought tolerance in rice. On the other hand, a calmodulin-like calcium signalling gene, OsCML16, was only increased the expression level in ‘4610’ in 14-day time point. It has been found that OsCML16 interacts with an ethylene-responsive element-binding factor, OsERF48, and enhances root growth and drought tolerance (Jung et al. 2017).

We further examined the expression patterns of TFs, known drought-tolerant genes and ROS scavengers. The expression profile showed that a number of TFs were upregulated in both ‘4610’ and Rondo, but most of them had relatively much higher expression in ‘4610’, especially at 14-day time point (Fig. 3). OsBZ8 (Os01g0658900) was the only TF that was differentially expressed in both genotypes and in both time points. Comparing to control groups, the log2 fold-change was 1.32 and 1.25 in ‘4610’, and 2.51 and 2.52 in Rondo, in earlier and later grain-filling stages, respectively. OsBZ8 is a G-box-binding factor-type bZIP protein that binds to ABA-responsive elements (ABREs) such as Rab16 and LEA genes, which regulate the drought stress response gene in the ABA-dependent pathway (Nakagawa et al. 1996). Other than OsBZ8, there were three other bZIP TFs that were differentially expressed under drought conditions. A TGA-type bZIP TFs (Os04g0637000) was exclusively expressed in Rondo. The Os04g0637000 homologous gene in Arabidopsis has been known to play a role in pathogen resistance and root hair development (Wang and Fobert 2013; Canales et al. 2017). Interestingly, the expression pattern of the other two bZIP TFs, Os01g0687600 and Os02g0766700, was similar in ‘4610’ and Rondo but at different time points. Both genes are related to ABA signalling under abiotic stress (Amir Hossain et al. 2010). However, they were upregulated at earlier time in ‘4610’ and later in Rondo (Fig. 3).

Several TFs shared the same trend that commonly upregulated in ‘4610’ at earlier stage and in Rondo at later stage, such as Os02g0649300 (HD-ZIP protein), Os3g0820500 (Actin depolymerizing factor) and Os03g0820300 (TFIIIA-type zinc finger protein). Some other TFs commonly increased in ‘4610’ in both time points but only in the later time point in Rondo, including the NAC TFs, Os11g0126900 and Os12g0123700. Both Os11g0126900 and Os12g0123700 were categorized into NAC group III, namely stress-responsive NAC genes (SNAC) in a previous study (Fang et al. 2008). Notably, there was one TF, WRKY12 (Os01g0624700), which was only upregulated at the earlier time point in ‘4610’ and had log2 fold-change over 6, which means the expression was 64 times higher under drought stress (Fig. 3). WRKY12 family proteins are a class of plant-specific TFs that involve in several stress response pathways (Liu et al. 2005; Ryu et al. 2006). WRKY family proteins that were reported previously in rice were mainly in pathogen defence regulation. However, several studies revealed that WRKY genes also play a role in enhancing tolerance to abiotic stresses such as drought, cold and salinity in soybean, wheat, Arabidopsis and rice (Zhou et al. 2008; Wu et al. 2009; Niu et al. 2012). Taken together, our data suggest that ‘4610’ might have triggered the signalling pathway earlier in response to drought stress than Rondo and had a better response outcome than Rondo accordingly.

Aside from ABA-induced gene families, DREB also regulates drought and cold stress response genes via the ABA-independent pathway (Shinozaki and Yamaguchi-Shinozaki 2000; Lata and Prasad 2011). The current transcription profile has a total of 19 annotated DREB family genes in rice. Our data showed that none of them were identified at the 14-day time point, and only four of them were identified solely at Rondo at the 21-day time point (Fig. 3). In contrast, ‘4610’ had many TFs turned on and off much earlier, including WRKY TF 12 (Os01g0624700) and bZIP TFs (Os01g0867300, Os02g0766700). This profile suggests that the stress-responsive reactions in Rondo had a distinct mechanism from ‘4610’, and the delayed and prolonged signalling might lead to overreacting response detected at the later time point.

Expression patterns of stress response genes

ROS scavengers enhanced drought tolerance. Reactive oxygen species scavengers play an important role in stress tolerance mechanism in plants (Das and Roychoudhury 2014). Under drought stress, the increased ROS accumulation (Mittler 2002) causes oxidative stress and damage in different levels such as proteins, lipids, DNA and RNA (Sgherri et al. 1993; Moran et al. 1994; Boo and Jung 1999). There are ROS scavenging mechanisms in plants that help protect the cells under drought stress, such as the scavenging enzymes and non-enzyme antioxidants. Two main enzymes are acting in the ascorbate/glutathione scavenging pathway, the APX and the glutathione reductase (GR), which convert toxic O$_2$- to H$_2$O (Foyer and Noctor 2011). Other ROS scavenging enzymes include superoxide dismutase (SOD), catalase (CAT), peroxidase (POD) and GSTs (Cruz de Carvalho 2008; Kumar and Trivedi 2018). There were eight ROS scavengers identified in the current study, including mitochondrial sulfur dioxygenase (Os0100667200), peroxidase 11 precursor (Os06g0274800), OsAPx7 (Os04g0948400) and GSTs (Os10g0529500, Os01g0682000, Os03g0785900, Os01g094750, Os10g0528100) (Fig. 3). Except for the peroxidase 11 precursor, which was solely upregulated in Rondo at the 14-day time point, the rest of ROS scavengers had relatively higher expression in ‘4610’, especially at the 14-day time point. Glutathione-S-transferases were the most abundant ROS scavengers in the current study. Although GST is not directly involved in the ascorbate/glutathione scavenging pathway, a previous study revealed that GST could protect the cell from oxidative damage by quenching reactive molecules with the addition of glutathione (Kumar and Trivedi 2018). Glutathione-S-transferases have also been proven to enhance chilling and drought tolerance in rice (Guo et al. 2006). Among the five GSTs identified, four of them were upregulated during earlier grain-filling stage in ‘4610’ and none were upregulated during this earlier stage in Rondo. Notably, our study did not identify any significant DEGs belong to the SOD and CAT families. These data again showed that most ROS scavengers responded much earlier and more upregulated in ‘4610’ than in Rondo. These ROS scavengers removed the excess ROS, relieve oxidative stress and subsequently prevent further damage. This factor may also partially contribute to the higher tolerance of ‘4610’ under drought than Rondo.

Chaperons induced under drought stress. We investigated the expression pattern of LEA genes, directly downstream genes of bZIP TFs. Late embryogenesis abundant protein helps offset the harmful effects of water deficit by preventing protein aggregation in the cell during water stress that allows the cell to maintain its function (Goyal et al. 2005). Therefore, LEA proteins were annotated as drought and cold tolerance proteins and have been overexpressed in various plants, including brassica and rice, to improve their tolerance to such stresses (Xiao et al. 2007; Dalal et al. 2009). In our study, eight LEA genes were differentially expressed; however, only one, i.e. Os01g0705200, was increased in both ‘4610’ and Rondo at the two time points. On the other hand, three of them were upregulated in ‘4610’ at the earlier
stage but in Rondo at the later stage, and four LEA genes were solely increased in Rondo at the later stage (Fig. 3). Our results clearly showed that at the earlier stage, LEA family proteins relatively had higher expression levels in ‘4610’ than Rondo. This result concurred with the expression pattern of TFs that has been discussed above, which suggests that ‘4610’ responds to the signalling cascades earlier in the grain-filling stage than Rondo, which may partially contribute to better drought tolerance.

Other genes contribute to drought tolerance. Several cytochrome p450 genes were upregulated in both stages in Rondo (Fig. 3). This may be the unique drought response strategy in Rondo. Previous studies reported that the expression of cytochrome p450 genes improves drought tolerance by fine-tuning GA-to-ABA balance and mediating metabolic responses (Nam et al. 2014; Tamiru et al. 2015) and grain formation during drought stress (Pandian et al. 2020). Several of these genes were upregulated in both time points of grain filling, some during the earlier stage, while others during the later stage. It was unexpected that these gene families were upregulated in Rondo, the more susceptible cultivar to drought. However, it is possible that the timing of the gene expression plays a significant role (most of the drought transcriptomic studies were performed during reproductive stage) and other gene networks could possibly mask the significant effects of the gene families. This phenomenon needs further investigation. However, at the same time, this also offers some opportunities for future molecular genetic manipulation to enhance drought tolerance of Rondo via modification of the key cytochrome p450 gene families.

Other than consistently expressed DEGs, there were several DEGs that were upregulated in ‘4610’ at the earlier stage but not in Rondo at the same stage (Fig. 3), including two-pore potassium channel, OsTPKb (Os07g0108800), Rab21 (Os11g0454300), Rab16 (Os11g0454400), OsIAA6 (Os01g741900), an auxin-responsive protein, and beta-carotene hydrolase (Os03g125100). Vacular two-pore potassium channels are important for the regulation of cellular potassium levels. Previous studies have shown that potassium uptake was greater in OsTPK-overexpressed rice lines. With higher cytoplasm to vacuole potassium ratio, plants tend to have better osmotic and drought tolerance (Ahmad et al. 2016); on the contrary, OsTPK knockout mutants showed lower drought tolerance (Ahmad et al. 2016; Chen et al. 2017). Both Rab16 and Rab21 are ABA-responsive genes, which can sense the existence of ABA trigger downstream stress response signalling. In addition, Rab21 was also known to increase rapidly after drought stress throughout all rice growth stages (Skriver et al. 1991; Yi et al. 2010). Rab proteins, the small G protein family, involved in different activities, including vesicle trafficking, intracellular signalling events, various physiological processes and stress response (Choudhary and Padaria 2015). The link between auxin expression and drought responses has been reported in a previous study with an example of OsIAA6. The study showed that OsIAA6-overexpressed rice had better drought tolerance, and it may be due to the control of tiller outgrowth (Jung et al. 2015). OsIAA and another gene, OsDHODH, which encodes cytosolic dihydroorotate dehydrogenase (DHODH) (Liu et al. 2009) were upregulated in both stages. OsDHODH is involved in plant stress response, including under salinity or drought stress conditions (Liu et al. 2009). Beta-carotene hydrolase is known to be involved in the synthesis of zeaxanthin, a carotenoid precursor of ABA biosynthesis. It has been reported that rice had significantly increased drought and oxidative stress tolerance with a higher level of zeaxanthin and ABA level (Du et al. 2010). Again, these data also suggest that ‘4610’ had a more rapid and better response against drought stress than Rondo.

**Yield-related performance under drought stress and its closely related gene regulation**

In agreement with previous studies, our results also showed that filled grain number, spikelet fertility, grain weight, plant dry weight, panicle length and yield were decreased under drought stress (Garrity and O’Bole 1994; van der Weijde et al. 2017; Zhang et al. 2018). Based on the DEGs and enrichment analysis, we identify several genes common in ‘4610’ and Rondo that may partially contribute to the phenotypic differences. For example, Os01g0327200, annotated as regulator of panicle-branching, grain weight, grain yield and photosynthesis; GWS (Os05g0158500), a serine carboxypeptidase, which acts as a positive regulator of grain size; cytokinin-activating enzyme (Os01g0588900) and cytokinin signalling kinase (Os02g0738400) were downregulated in both ‘4610’ and Rondo in the 14-day time point but not in the 21-day time point [see Supporting Information—Table S1]. Conversely, during the later grain-filling stage, there were no common downregulated DEGs across the two genotypes that directly related to agronomic traits, but there were eight downregulated DEGs that encoded photosystem proteins (Os01g0773700, Os03g0747700, Os03g0778100, Os07g0148900, Os07g0673550, Os08g0119800, Os09g0475800 and Os09g0481200) [see Supporting Information—Table S2]. Photosynthesis provides a carbon source for starch accumulation during the grain-filling stage in cereals (Prathap et al. 2019). Therefore, a decreased photosynthesis may contribute to poor grain filling, which affects the size and quality of the rice grains. In addition, some grain and yield-related genes were downregulated in Rondo, including regulation of grain size (Os06g0130400, Os09g0517600, Os12g0610200, Os04g0645100), regulation of grain shape (Os09g0448500) and regulation of yield (Os1g0878400, Os07g0603800) [see Supporting Information—Table S2]. Therefore, our study suggests that the inferiority of agronomic trait parameters under drought stress conditions might be partially contributed by the downregulation of both groups of trait-related genes and photosynthesis-related genes.

**Potential novel drought-responsive genes**

To explore the potential novel drought-responsive genes, we examined the DEGs without functional annotation. During the earlier time point, 59 DEGs annotated as ‘conserved hypothetical gene’, 45 DEGs annotated as ‘hypothetical gene/protein’ and 23 DEGs annotated as ‘domain of unknown function (DUF) proteins’ in ‘4610’; on the other hand, 133 DEGs annotated as ‘conserved hypothetical gene’, 34 DEGs annotated as ‘hypothetical gene/protein’ and 15 DEGs annotated as ‘DUF proteins’ in Rondo [see Supporting Information—Table S1]. During the later stage, 29 DEGs annotated as ‘conserved hypothetical gene’, 25 DEGs annotated as ‘hypothetical gene/protein’ and 6 DEGs annotated as ‘DUF proteins’ in ‘4610’; on the other hand, 326 DEGs annotated as ‘conserved hypothetical gene’, 168 DEGs annotated as ‘hypothetical gene/protein’ and 107 DEGs annotated as ‘DUF proteins’ in Rondo [see Supporting Information—Table S2]. To narrow down our potential novel gene pool, three screening criteria were developed: (i) DEGs that had log2 fold-change larger than 1 and consistently upregulated in both ‘4610’ and Rondo in the same stage, (ii) DEGs in only one genotype but consistently upregulated in both stages with log2 fold-change larger than 1 and (iii) DEGs consistently in both genotypes and both stages. A total of 36 DEGs met our criteria.
with log2 fold-change ranged from 0.72 to 7.54 (Table 4). Two DEGs that were upregulated in all four groups (Os02g0463401 and Os11g0533400), and in the case of these two genes, the log2 fold-change in ‘4610’ during earlier stage was the highest (2.04 and 3.07, respectively). Four DEGs were upregulated only in ‘4610’ during both stages (Os01g0184050, Os02g0259900, Os03g0267100 and Os08g0110600). There were also several DEGs that were upregulated in both genotypes during grain filling, and one of the genes with the largest log2 fold-change was Os02g0609000 (6.04 and 7.54 in ‘4610’ and Rondo, respectively). Some of these genes may have functions related to drought stress tolerance; however, this requires further investigation.

**Conclusions**

This transcriptomics study showed the complexity of the drought response mechanism in rice during grain filling. Phenotypic and transcriptomic data showed that both ‘4610’

| Gene ID       | Annotation                                      | Reproductive stage | Grain-filling stage |
|---------------|-------------------------------------------------|--------------------|--------------------|
| Os01g0128250  | Hypothetical gene                               | ns                 | 2.35               |
| Os01g0184050  | Hypothetical protein                            | 1.34               | 1.62               |
| Os01g0200350  | Hypothetical protein                            | 2.36               | 1.51               |
| Os01g0214500  | Conserved hypothetical protein                  | 1.66               | 1.07               |
| Os01g0229600  | Conserved hypothetical protein                  | 1.77               | 1.18               |
| Os01g0652375  | Hypothetical protein                            | ns                 | 1.61               |
| Os01g0652450  | Hypothetical gene                               | ns                 | 5.46               |
| Os01g0727700  | Hypothetical conserved gene                     | 1.92               | 2.30               |
| Os01g0727820  | Hypothetical protein                            | ns                 | 1.78               |
| Os01g0838350  | Conserved hypothetical protein                  | ns                 | 1.12               |
| Os01g0888900  | Conserved hypothetical protein                  | ns                 | 2.00               |
| Os02g0140800  | Conserved hypothetical protein                  | 2.12               | 1.28               |
| Os02g0258800  | Conserved hypothetical protein                  | 1.77               | 2.29               |
| Os02g0259900  | Conserved hypothetical protein                  | 1.04               | 1.08               |
| Os02g0463401  | Conserved hypothetical protein                  | 2.04               | 0.72               |
| Os02g0463401  | Conserved hypothetical protein                  | ns                 | 2.09               |
| Os02g0745500  | Hypothetical protein                            | ns                 | 2.43               |
| Os02g0757700  | Hypothetical protein                            | ns                 | 6.01               |
| Os02g0766700  | Hypothetical protein                            | 1.12               | 1.08               |
| Os03g0305550  | Hypothetical gene                               | 2.45               | 2.43               |
| Os03g0381500  | Conserved hypothetical protein                  | 1.96               | 1.36               |
| Os03g0809000  | Hypothetical conserved gene                     | ns                 | 1.17               |
| Os05g0299500  | Protein of unknown function DUF914, eukaryotic family protein | ns | 1.23 | 1.55 |
| Os05g0390550  | Conserved hypothetical protein                  | 1.92               | 2.71               |
| Os06g0651200  | Conserved hypothetical protein                  | 2.26               | 0.88               |
| Os06g0706500  | Hypothetical protein                            | 1.22               | 2.30               |
| Os07g0567200  | Conserved hypothetical protein                  | 1.07               | 1.9                |
| Os08g0111600  | Protein of unknown function DUF1442 domain-containing protein | 2.27 | ns | 1.25 |
| Os08g0285500  | Hypothetical conserved gene                     | ns                 | 1.74               |
| Os09g0425000  | Hypothetical protein                            | 1.9                | 1.29               |
| Os09g0426000  | Protein of unknown function DUF6, transmembrane domain-containing protein | ns | 1.65 | 2.65 |
| Os11g0533400  | Conserved hypothetical protein                  | 3.07               | 1.27               |
and Rondo were affected by drought stress. Both the moderate-tolerant and susceptible genotypes had variable responses against drought stress, including drought sensing, signalling, downstream regulation and ROS scavenging. By comparing the expression of these key genes, we suggest that ‘4610’ was less affected by drought stress due to its more rapid stress response and higher expression level of key drought-tolerant genes, LEA proteins, ROS scavengers, APXs and GSTs. Some of these genes can be potential targets for further study and manipulation to develop more resilient high-yielding rice under drought stress conditions.

**Supporting Information**

The following additional information is available in the online version of this article—

**Table S1.** Differentially expressed genes at the earlier grain-filling stage (Day 14).

**Table S2.** Differentially expressed genes at the later grain-filling stage (Day 21).

**Table S3.** Gene ontology at the earlier grain-filling stage (Day 14).

**Table S4.** Gene ontology at the later grain-filling stage (Day 21).

**Table S5.** Differentially expressed genes at both time points of the grain-filling stage (Day 14 and Day 21).

**Figure S1.** Principal component analysis (PCA) of RNA-seq samples separated by the two time points (Day 14 vs. Day 21).

**Figure S2.** Venn diagrams of differentially expressed genes (DEGs) for each genotype.

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**Contributions by the Authors**

Y.L., R.T., L.T. and E.S. designed the experiment; E.S. conceived the project; Y.L., R.T., A.M., L.T. and E.S. performed the experiment; Y.L. analyzed data; E.S. supervised the project; Y.L. wrote the first draft of the manuscript; E.S. and L.T. reviewed and edited the manuscript; all read and approved the manuscript.

**Conflicts of Interest**

The authors declare that there is no conflict of interest.

**Data Availability**

Data are available as Supporting Information.

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