CG and CHG Methylation Contribute to the Transcriptional Control of OsPRR37-Output Genes in Rice

Chuan Liu*1, Na Li1, Zeping Lu1, Qianxi Sun1, Xinhan Pang1, Xudong Xiang1, Changhao Deng1, Zhengshuo Xiang1, Kunxian Shu1, Fang Yang2 and Zhongli Hu2

1 Chongqing Key Laboratory of Big Data for Bio Intelligence, Chongqing University of Posts and Telecommunications, Chongqing, China, 2 State Key Laboratory of Hybrid Rice, College of Life Sciences, Wuhan University, Wuhan, China

Plant circadian clock coordinates endogenous transcriptional rhythms with diurnal changes of environmental cues. OsPRR37, a negative component in the rice circadian clock, reportedly regulates transcriptome rhythms, and agronomically important traits. However, the underlying regulatory mechanisms of OsPRR37-output genes remain largely unknown. In this study, whole genome bisulfite sequencing and high-throughput RNA sequencing were applied to verify the role of DNA methylation in the transcriptional control of OsPRR37-output genes. We found that the overexpression of OsPRR37 suppressed rice growth and altered cytosine methylations in CG and CHG sequence contexts in but not the CHH context (H represents A, T, or C). In total, 35 overlapping genes were identified, and 25 of them showed negative correlation between the methylation level and gene expression. The promoter of the hexokinase gene OsHXK1 was hypomethylated at both CG and CHG sites, and the expression of OsHXK1 was significantly increased. Meanwhile, the leaf starch content was consistently lower in OsPRR37 overexpression lines than in the recipient parent Guangluai 4. Further analysis with published data of time-course transcriptomes revealed that most overlapping genes showed peak expression phases from dusk to dawn. The genes involved in DNA methylation, methylation maintenance, and DNA demethylation were found to be actively expressed around dusk. A DNA glycosylase, namely ROS1A/DNG702, was probably the upstream candidate that demethylated the promoter of OsHXK1. Taken together, our results revealed that CG and CHG methylation contribute to the transcriptional regulation of OsPRR37-output genes, and hypomethylation of OsHXK1 leads to decreased starch content and reduced plant growth in rice.

Keywords: rice growth, DNA methylation, RNA-seq, circadian clock, output genes

INTRODUCTION

Plant DNA methylation occurs in the sequence context of CG, CHG, and CHH (where H is A, C, or T) (Zhang et al., 2006). DNA methylation is considered a stable epigenetic mark that can be transmitted across generations (Gehring, 2019). A specific DNA methylation state is also under the dynamic regulation by de novo methylation, maintenance of methylation, and active demethylation (Law and Jacobsen, 2010; Zhang et al., 2018). The function of DNA methylation in plant reproductive cells has been extensively studied, and this knowledge has deepened our
understanding of the dynamic DNA methylation patterns during plant development (Zheng et al., 2019; Higo et al., 2020; Kim et al., 2021; Zhou et al., 2021a). In addition, DNA methylation reportedly has roles in regulating plant architecture, plant defense against rice black-streaked dwarf virus, and multiple agronomical traits (Zhang et al., 2015, 2020; Kawakatsu, 2020; Xu et al., 2020). However, whether DNA methylation plays a role in regulating circadian clock output genes remains unclear.

Circadian clock comprises multiple transcription-translation feedback loops, which function to improve plant environmental adaptation. Altered expression of circadian clock genes, such as CIRCADIAN CLOCK ASSOCIATED1 (CCA1), can increase levels of plant growth and fitness (Dodd et al., 2005; Masuda et al., 2020). Meanwhile, expression amplitude of CCA1 is associated with the CHH methylation level in the promoter region, which determines plant growth vigor (Ng et al., 2014). Recently, it was reported that the circadian clock genes ZEITLUPE and TIMING OF CAB EXPRESSION 1 act downstream of DNA methyltransferases to control circadian rhythm (Tian et al., 2021). These results shed light on DNA methylation-mediated regulation of clock gene expression. Pseudo-Response Regulators (PRRs) are key components of transcription-translation feedback loops in plants and mediate the circadian regulation of clock output genes (Farre and Liu, 2013), including genes involved in the regulation of growth, flowering, abiotic stress, and yield-related traits (Li C. et al., 2020; Li N. et al., 2020; Sun et al., 2020; Wei et al., 2020; Liang et al., 2021). OsPRR37 was primarily identified to delay the flowering time and increase grain yield and adaptation in rice (Koo et al., 2013; Liu et al., 2013, 2015; Yan et al., 2013; Gao et al., 2014; Fujino et al., 2019). A recent study further revealed a distinct role of OsPRR37 in promoting flowering in the japonica variety Zhonghua 11 under natural long-day conditions (Hu et al., 2021). Although several output genes involved in the photoperiodic flowering pathway are used to explain the trait variations (Chen et al., 2021; Zhou et al., 2021a), the underlying mechanism of how OsPRR37 regulates its output genes and multiple traits remains unclear.

Circadian regulation of plant transcriptome benefits the acute responses of plants to the daily fluctuating environment (Panter et al., 2019). Our previous study confirmed that OsPRR37 protein functions as a transcriptional repressor and confers expanded regulation of transcriptome rhythms (Liu C. et al., 2018). The regulation of circadian-regulated genes by DNA methylation in Populas trichocarpa suggested that DNA methylation contributes to the expression levels of clock output genes (Liang et al., 2019). Based on these results, we hypothesize that OsPRR37 uses DNA methylation as a pathway to regulate the transcription of its output genes. In the present study, we sought to determine whether and how DNA methylation regulates OsPRR37-output genes. To this end, whole-genome bisulfide sequencing (WGBS) and high-throughput RNA sequencing (RNA-seq) were applied to identify the overlapping genes, which were considered to be the OsPRR37-output genes regulated by DNA methylation. The available data of time-course transcriptomes were used to confirm the expression change of overlapping genes and to analyze the genes involved in DNA methylation pathways. Our results revealed that DNA methylation was an alternative medium for OsPRR37 to regulate the output genes and plant growth.

MATERIALS AND METHODS

Plant Materials and Growth Conditions

The OsPRR37 overexpression lines (OE5 and OE9) were described as previously reported (Liu C. et al., 2018). Briefly, OsPRR37 overexpression lines were generated by overexpressing OsPRR37 in an elite rice variety, namely Guangluai 4 (GL). NIL-OsPRR37 is a nearly isogenic line in the GL background and contains the functional allele of OsPRR37 from the elite variety Teqing. Rice growth phenotypes were obtained from the plants growing under natural long-day conditions in Wuhan University, Wuhan, China (30°54’01”N, 114°37’23”E). For WGBS and RNA sequencing, seeds of GL and OE5 were planted in a growth chamber (PRX-380B, Shanghai Guning Instrument Co., Ltd) for 15 days after germination. The growth chamber was set at 28°C under a 14-h light/10-h dark cycle with the light period of 6:00–20:00. The top most expanded leaves were harvested at 9:00, frozen in liquid nitrogen, and then stored at −80°C for DNA and RNA extraction. The same two biological replicates were applied to WGBS and RNA-seq.

Bisulfite-Seq Library Generation and Sequencing

Briefly, total genomic DNA of rice leaves was extracted using the cetyltrimethylammonium bromide method (Doyle, 1987). The DNA concentration and quality were estimated using NanoDrop 2000 spectrophotometer (NanoDrop Technologies, Wilmington, DE, United States), Qubit 3.0 fluorometer (Life Technologies, Carlsbad, CA, United States), and 1.0% agarose gel electrophoresis. Then, 2-µg genomic DNA spiked with 5-ng unmethylated Lambda DNA (Promega, Madison, WI, United States) was fragmented by sonication to generate fragments measuring 300–500 bp. These fragments were then ligated with 5-methylcytosine-modified adapters and subjected to bisulfide conversion using the ZYMOS ZY DNA Methylation-Gold Kit (Zymo Research, Irvine, CA, United States). The bisulfide-converted DNA was purified, recycled, and then amplified by PCR with 10 cycles using KAPA HotStart ReadyMix (Kapa Biosystems, Wilmington, MA, United States) and Illumina 8-bp index primers. The WGBS libraries were analyzed using the Bioanalyzer 2100 system (Agilent Technologies, CA, United States) and sequenced on Illumina NovaSeq 6000 with a paired-end sequencing length of 150 bp (PE150) at Frasergen Bioinformatics Co., Ltd (Wuhan, China). The percentage of cytosines sequenced at cytosine reference positions in the lambda genome was considered to reflect the overall sodium bisulfite non-conversion rate.

Bisulfite-Seq Data Processing and Analysis

Quality control of WGBS data was performed using FastQC (version 0.11.9, Babraham Bioinformatics, United Kingdom).
Sequencing adapters and low-quality reads were removed using Trimmomatic (Bolger et al., 2014). The trimmed reads were then aligned and mapped to the rice reference genome of Nipponbare (MSU_v7.0) using Bismark (Krueger and Andrews, 2011). The percentage methylation level was calculated by mC/(mC + umC), where mC and umC represent the number of methylated and unmethylated reads, respectively. Only the CG/CHG/CHH sites with a read coverage of ≥ 5 across all samples were used for differential methylation analyses. Differentially methylated regions (DMRs) were identified using the R package “dmrseq” (Korthauer et al., 2019). Regions with a q-value < 0.05, number of CG/CHG/CHH sites ≥ 5 and methylation difference > 20% were defined as DMRs. DMR distribution on rice chromosomes was plotted using Circos (version 0.69) (Krzywinski et al., 2009). DMRs were annotated using ChiPseeker package (Yu et al., 2015). Methylation status along the genomic regions of DMRs ± 20 kb was plotted using pyGenomeTracks (version 3.6) (Lopez-Delisle et al., 2021).

RNA Library Generation and Sequencing
Total RNA from rice leaves was extracted using TRIzol Reagent (Invitrogen, CA, United States) for RNA sequencing. RNA purity and integrity were analyzed using a NanoDrop 2000 spectrophotometer (NanoDrop Technologies, Wilmington, DE, United States) and the Bioanalyzer 2100 system (Agilent Technologies, CA, United States). RNA contamination was assessed by 1.5% agarose gel electrophoresis. A total of 1 µg of RNA per sample was used as the input material for library preparation. The mRNA was purified from the total RNA using poly-T oligo-attached magnetic beads. Sequencing libraries were generated from the purified mRNA using the VAHTS Universal V6 RNA-seq Library Kit for MGI (Vazyme, Nanjing, China) following the manufacturer’s recommendations with unique index codes. The library quantification and size were assessed using a Qubit 3.0 fluorometer (Life Technologies, Carlsbad, CA, United States) and the Bioanalyzer 2100 system (Agilent Technologies, CA, United States). Subsequently, sequencing with a paired-end sequencing length of 150 bp (PE150) was performed on the MGI-SEQ 2000 platform (MGI Tech Co., Ltd. Shenzhen, China) by Frasergen Bioinformatics Co., Ltd (Wuhan, China).

Bioinformatics Analysis of RNA Sequencing and Microarray Data
Sequencing adapters and low-quality reads were removed with fastp (version 0.20.1) (Chen et al., 2018), and the quality of raw reads was evaluated with FastQC (version 0.11.9, Babraham Bioinformatics, United Kingdom). The remaining clean reads were mapped to the rice reference genome of Nipponbare (MSU_v7.0) using Hisat2 (version 2.1.0) (Kim et al., 2019). Mapping statistics were generated using Samtools (version 1.11) (Li H. et al., 2009). TPMCalculator was used to count the reads mapped to individual genes as well as to measure gene expression levels by calculating transcripts per million (TPM) read values (Vera Alvarez et al., 2019). Differentially expressed genes (DEGs) were identified using DESeq2 (Love et al., 2014). Gene Ontology (GO) enrichment analysis was performed using the GO annotation file MSU7.0 gene ID (TIGR) of agrigo v2.0 and clusterProfiler 4.0 (Tian et al., 2017; Wu et al., 2021). KEGG enrichment analysis was performed using KOBAS 3.0, and the output data were plotted using clusterProfiler 4.0 (Bu et al., 2021). Gene symbols with known or unknown function were annotated using MBKbase-rice database1, funRiceGenes database2, and China Rice Data Center3 (Yao et al., 2018; Peng et al., 2020). The raw RNA-seq data of time-course transcriptomes, which were downloaded from NCBI-GEO database (GSE114188), were reanalyzed as per the RNA-seq data processing pipeline in this study. The corresponding time-course samples of GL and OE5 comprise six time points (4:00, 8:00, 12:00, 16:00, 20:00, and 0:00) with three replicates at 45 days growth under natural long-day conditions. Statistical significance of different expressions was evaluated by unpaired Student’s t-test at each time point. The microarray data were obtained by GSE19024 on NCBI-GEO (Wang et al., 2010). The corresponding tissue samples of interest were described in Supplementary Table 1, which were the subset of samples in a previous study (Wang et al., 2010). Before being used to plot the heatmap, the signal values of biological and technical replicates for the same tissue were averaged.

Quantitative RT-PCR and Starch Content Determination
Quantitative RT-PCR was conducted with the same protocol as previously reported (Liu et al., 2015). The PCR primer sequences were 5’-TGACAAAGCCTAGTACAAATGAGAGAG-3’ and 5’-CAGTGCGTGCGCAGGATGAAAATG-3’. Approximately, 0.2 g of fresh leaf samples were weighed before estimating the starch content. Starch content was determined according to previously published protocols (Smith and Zeeman, 2006).

RESULTS

Rice Growth Was Repressed by Overexpressing OsPRR37
During the field trials, we observed that rice growth was retarded in OsPRR37 overexpression lines (OE). To investigate the effects of OsPRR37 overexpression on rice growth, we recorded the morphology and dry weight of GL, OE5, OE9, and NIL-OsPRR37 at 25, 40, and 55 days after sowing the seeds. The growth of OE5 and OE9 was significantly repressed compared to the growth of GL and NIL-OsPRR37 on these days (Figures 1A–F). These results suggest that natural loss-of-function and gain-of-function alleles of OsPRR37 showed a comparable growth rate during the vegetative growth period. Then, the diurnal expression profile of OsPRR37 was monitored over a day using quantitative RT-PCR. OsPRR37 was identified as having similar expression rhythms in GL and NIL-OsPRR37 as well as their peak expression phase was around 12:00. Conversely, the expression of OsPRR37 in OE5 and OE9 was much higher and showed altered rhythms with the peak expression phase around 4:00 (Figure 1G).

1 http://www.mbkbase.org/rice/
2 https://funricegenes.github.io/
3 https://www.ricedata.cn/gene/
results suggested that the overexpression of OsPRR37 changed its diurnal rhythms and repressed rice growth. In a previous study, it was reported that OsPRR37 widely regulates output genes and particularly suppresses output genes with phases around 9:00 (Liu C. et al., 2018). To investigate whether DNA methylation associated with OsPRR37 regulates the output genes, samples of GL and OE5 at 9:00 were subjected to WGBS and RNA-seq. The workflow of this study is shown in Figure 1H.

Whole-Genome Bisulfide Sequencing and RNA Sequencing Quality Assessment and Alignment
Whole-genome bisulfide sequencing and RNA-seq were used to investigate the role of DNA methylation in the transcriptional regulation of OsPRR37-output genes. WGBS generated 48,752,185 and 63,773,173 raw reads for the two GL replicates and 59,530,161 and 56,527,166 raw reads for the two OE5 replicates. After quality control filtration, 45,246,229 and 59,675,582 clean reads remained for OE5. The clean reads ratio ranged from 92.8 to 94.2%. The average percentage of Q30 and GC content for the four sequencing libraries was 92.7 and 22.9%, respectively (Supplementary Table 2). Of those clean data, 53.4% (GL-1), 51.7% (GL-2), 54.2% (OE5-1), and 52.2% (OE5-2) were uniquely mapped to the rice genome (Supplementary Table 3). Overall, 30,872,222 CG sites, 27,422,379 CHG sites, and 104,533,760 CHH sites were identified with sequencing coverage between 84.9 and 86.1%. Of these clean read pairs, 90.0 to 90.7% were mapped to RNA-seq to profile the genome-wide gene expressions. The correlation between samples was ranged from 0.9804 to 0.9929 (Supplementary Figure 1), whereas the expression correlation between samples was ranged from 0.9804 to 0.9929 (Supplementary Figure 2).

Overexpressing OsPRR37 Altered Global CG and CHG Methylation
To identify DMRs, three cytosine sequence contexts (CG, CHG, and CHH) were separately applied to differential methylation analysis. Genomic regions with a q-value of < 0.05 and a differential methylation level of > 20% were considered as DMRs. A total of 321 and 949 DMRs were found in CG (DMR-CG) and CHG (DMR-CHG) sequence contexts, respectively. However, no DMRs were identified in the CHH sequence context. Among DMR-CG, 90 were hypermethylated and 231 were hypomethylated (Figure 2A). Conversely, among DMR-CHG, 480 were hypermethylated and 469 were hypomethylated (Figure 2B). These data revealed a higher proportion of hypomethylated DMR-CG (72.0%) than DMR-CHG (49.4%). Furthermore, the significance of methylated DMRs across the 12 chromosomes found that DMRs were evenly distributed on the rice genome and were of high significance (Figures 2C,D).

To obtain DMR-associated genes (DMGs), DMR-CG and DMR-CHG were both annotated with the R package “ChIPseeker.” The result showed that 32.7%, 20.6%, and 15.9% of DMR-CG were located in the promoter region at ≤ 1 kb, 1–2 kb, and 2–3 kb upstream of transcription start site, respectively (Figure 3A). The distal intergenic region accounted for 20.2% of DMR-CG. Conversely, only a small fraction of DMR-CG was annotated within Intron (2.5%), Exon (2.5%), Downstream (≤ 1 kb: 0.9%, 1–2 kb: 2.2%, and 2–3 kb: 1.2%) and 3′UTR (1.2%). This means that most DMR-CG (69.2%) were located in the promoter region (Figure 3A). Similarly, 70.3% of DMR-CHG were located in promoter region (Figure 3B). These results supported the notion that cytosine methylation majorly occurred in the promoter region. Then, we performed functional enrichment analysis with these DMGs. The network of five most enriched GO ontologies for DMR-CG-associated genes showed that LOC_Os02g03540 and LOC_Os06g14360 enable ribose phosphate diphosphokinase activity and are involved in ribonucleoside monophosphate biosynthetic process (Figure 3C). LOC_Os03g45410/OsTBP2 and LOC_Os03g14720 enable obsolete RNA polymerase II transcription factor activity and are involved in transcription initiation from the RNA polymerase II promoter. LOC_Os03g45410/OsTBP2 was reported to be a TATA-binding protein, which interacts with the transcription factor IIB (Zhu et al., 2002). Interestingly, LOC_Os03g14720 is a putative transcription initiation factor IIF. These results highlighted that DMR-CG-associated genes mainly function in gene transcription regulation. However, no GO term was found to be significantly enriched for DMR-CHG-associated genes.

Differentially Expressed Gene Analysis
Although OsPRR37 overexpression altered the methylation of > 1,000 DMGs, the number of transcriptionally regulated DMGs remains unknown. Samples of GL and OE5 were subjected to RNA-seq to profile the genome-wide gene expressions. The overall gene expression level was slightly higher for OE5 than for GL (Supplementary Figure 1), whereas the expression correlation between samples was ranged from 0.9804 to 0.9929 (Supplementary Figure 2). Genes with low expression (sum of TPM being < 2 in both GL and OE5) were filtered out. A total of 743 DEGs ([log2FC] > |log21.5|, adjusted P-value < 0.05) were identified between GL and OE5 (Supplementary Figure 3). Among these DEGs, 286 (38.5%) were downregulated and 457 (61.5%) were upregulated (Figures 4A,B). The increment of the mean TPM for upregulated DEGs was larger than the decrement of the mean TPM for downregulated DEGs (Figure 4A). The expression of DEGs in the two replicates was similar so that DEGs are robust to be further analyzed (Figure 4B).

GO enrichment analysis found that DEGs are instrumental in metal ion binding, transporter activity, and chitinase activity and are mainly involved in carbohydrate metabolic process, chitin catabolic process, defense response to bacterium and fungus, and nitrate assimilation, among other functions (Figures 4C,D). The KEGG pathway enrichment analysis showed that upregulated DEGs participate in amino sugar and nucleotide sugar metabolism, carbon metabolism, glycerolipid metabolism, nucleotide transport, and starch and sucrose metabolism. Conversely, downregulated DEGs were mainly involved in nitrogen metabolism, amino acid metabolism, post-translational modification, protein modification, and glycosaminoglycan biosynthesis.
Characterization of Overlapping Genes Between DMR-Associated Genes and Differentially Expressed Genes

Overlap analysis between DMGs and DEGs was performed to identify transcriptionally regulated DMGs. In total, 35 genes were found to be shared between DMG and DEG sets. Among these, five DEGs were found to be differentially methylated in both CG and CHG sequence contexts, whereas 19 DEGs were uniquely shared with DMR-CHG and 11 DEGs were uniquely shared with DMR-CG (Figure 5A). The correlation between methylation difference and expression fold-change of overlapping genes was investigated. Consequently, 14 of 16 genes (87.5%) showed negative correlation between expression and CG methylation, and 14 of 16 genes (87.5%) were methylated in promoter regions (Figure 5B). In contrast, 16 out of 24 genes (66.7%) showed a negative correlation between expression and CHG methylation, and 15 of 24 genes (62.5%) were methylated in promoter regions (Figure 5C). After removing the redundant genes, in total, 25 genes showed negative correlation between expression and cytosine methylation level (Figures 5B,C).

Functional annotation with the MBKbase, funRiceGenes database and China Rice Data Center identified seven genes with known function: OsHXK1 (LOC_Os07g26540), OsZIP9 (LOC_Os05g39540), SDT/OsmIR156h (LOC_Os06g44034), OsMADS18 (LOC_Os07g41370), OsPT11 (LOC_Os01g46860), OsRLCK109/OsBBS1 (LOC_Os03g24930), and OsNAS3 (LOC_Os07g48980). Among these genes, OsHXK1 showed the highest negative correlation between methylation difference...
FIGURE 2 | Identification and analysis of differentially methylated regions. Comparisons of hypermethylation and hypomethylation levels in GL and OE5 were plotted for both CG (A) and CHG (B) sequence contexts. Different letters above violin plots represent significant differences at $P < 0.01$ as revealed by one-way ANOVA analysis (Tukey’s multiple comparison test). DMR distribution on rice chromosomes in CG (C) and CHG (D) sequence contexts is shown. From outer to inner layers, the circular plots represent chromosomes, hyper-DMR distribution (the more outward means higher significance), heatmap of GC content (deeper red colors indicate higher GC content), heatmap of gene density (deeper green colors indicate higher gene density), and hypo-DMR distribution (greater proximity to the center of the circle indicates higher significance).

and expression fold-change (Figures 5B,C). Detailed methylation status indicated that the CG and CHG sites in the promoter region of OsHXK1 were both hypomethylated (Figure 5D).

Diurnal Rhythms and Functional Characterization of Overlapping Genes
As OsPRR37 is a key component in the rice circadian clock, diurnal rhythms of overlapping genes were further investigated with the reported time-course RNA-seq data of leaf samples at 45 days growth (Liu C. et al., 2018). Among the 35 overlapping genes, 29 were observed to be differentially expressed with at least one timepoint, which confirmed the identification of DEGs in this study (Figure 6A and Supplementary Figure 4). We also found that 31 overlapping genes showed diurnal rhythms, and 27 were observed to show a peak expression phase of 16:00–04:00. These data indicated that most differentially methylated DEGs were under circadian control and tended to function from dusk to dawn. Further investigation of the seven reported overlapping genes revealed that four of
them showed significant differences in their expression across different time points of the day (Figure 6A). Interestingly, OsHXK1, SDT/OsmiR156h, OsMADS18, and OsPT11 were diurnally expressed and showed a peak expression phase of 20:00–4:00, indicating that they predominantly function during the night. SDT/OsmiR156h can modulate the rice yield, plant architecture, and seed dormancy by targeting Ideal Plant Architecture1 (IPA1) (Jiao et al., 2010; Miao et al., 2019). Its continuously high expression suggested the involvement of the SDT/OsmiR156h-IPA1 module in OsPRR37-mediated rice growth regulation. Except for SDT/OsmiR156h, the other six genes were mapped in the microarray data of Zhenshan97 tissues (Wang et al., 2010). OsMADS18 was widely expressed in the tissues of seedling, leaf, shoot, sheath, stem, and panicle, which is in line with its function in flowering signal transduction (Figure 6B; Fornara et al., 2004; Yin et al., 2019). The significant repression of OsMADS18 can partly explain the delayed growth and flowering (Figure 6A). OsNAS3 encodes a nicotianamine synthase that is important for Fe homeostasis (Aung et al., 2019). Our result showed that OsNAS3 was widely expressed in germinating seed, plumule, radicle, seedling, leaf, root, shoot, sheath, stem, panicle, and spikelet. OsRLCK109/OsBBS1 was diurnally expressed with a peak phase of 16:00–20:00 and was highly expressed in leaf, root and sheath to regulate leaf senescence and salt stress responses (Zeng et al., 2018). OsZIP9 was mainly expressed in the root and sheath to uptake zinc for rice growth (Tan et al., 2020; Yang et al., 2020). However, even though OsPT11 is a rice phosphate transporter that regulates phosphate uptake and transport (Paszkowski et al., 2002; Yang et al., 2012), it was highly expressed in many tissues, such as germinating seed, radicle, root, leaf, sheath, stamen, endosperm, and panicle. The role of OsZIP9 and OsPT11 in coordinating ion uptake and rice growth needs to be further confirmed.
FIGURE 4 | Expression and functional enrichment of differentially expressed genes. (A) A comparison of mean expression levels of upregulated, no significant difference, and downregulated genes between GL and OE. Different letters above violin plots represent significant differences at $P < 0.01$ as revealed by one-way ANOVA analysis (Tukey’s multiple comparison test). (B) The heatmap of 457 upregulated and 286 downregulated DEGs. The TPM value of DEG was scaled by row with the “pheatmap” package in R. Dotplots of significant GO terms for upregulated (C) and downregulated (D) DEGs. Dotplots of significant KEGG pathways for upregulated (E) and downregulated (F) DEGs. GO terms and KEGG pathways with adjusted $P$-value $< 0.05$ were considered as significant, and if the number of significant terms or pathways was $> 15$, only 15 terms or pathways were plotted.
OsHXK1 was identified to be highly expressed in the germinating seed, leaf, stem, stamen, and endosperm (Figure 6B). This expression pattern is in close agreement with a previous study wherein OsHXK1 was reported to regulate reactive oxygen species in rice anthers (Zheng et al., 2019) and knockout of OsHXK1 improved rice photosynthetic efficiency and yield (Zheng et al., 2021). In plant leaves, starch is accumulated during the day and consumed by respiration at night, and therefore, the starch content can indicate the strength of photosynthesis. To investigate whether the photosynthesis product was altered by the significantly elevated expression of OsHXK1, we compared the total starch content in GL, OE5, OE9, and NIL-OsPRR37 leaves at the ending (20:00) and beginning (4:00) of the day. We found the total starch content to be significantly lower in OE5 and OE9 than in GL and NIL-OsPRR37 at both time points (Figure 6C). The low starch content in OE lines resulted in energy deficit, consequently causing repressed rice growth (Figure 1). These results revealed that the enhanced expression of OsHXK1 by hypomethylation decreased starch content and rice growth, thus suggesting that OsHXK1 is a key output gene applied by OsPRR37 to regulate rice growth.

**Diurnal Expression Analysis of DNA Methylation Related Genes**

DNA methylation patterns are decided by coordinated regulation of DNA methylation and demethylation pathways. To explore
the upstream genes of DMGs, the genes involved in DNA methylation were profiled with time-course RNA-seq data. In total, based on a recent study, 36 genes were grouped into the RNA-directed DNA methylation (RdDM) pathway, methylation maintenance, and DNA demethylation (Sun et al., 2021). Interestingly, most of them (26 genes) exhibited a peak expression phase of 16:00–20:00 (Figure 7 and Supplementary Figure 5), indicating that DNA methylation, methylation maintenance, and DNA demethylation are particularly active around dusk. Among these, several genes encoding Argonaute (AGO) proteins are altered, including the downregulated AGO1a and AGO1d and the upregulated AGO1b and AGO18 (Figure 7). The miR168-AGO1 module can regulate multiple miRNAs to improve yield, reduce flowering time, and enhance immunity (Wang et al., 2021). Meanwhile, AGO18 sequesters miR168 to alleviate the repression of rice AGO1 (Wu et al., 2015), and a regulation module of miR168a–OsAGO1/OsAGO18–miRNAs-target genes was proposed to regulate agronomically important traits (Zhou et al., 2021a). These results combined with our data suggest a causal link between rice growth repression and the altered module of miR168a-OsAGO1/OsAGO18.

A relatively similar expression of methylation maintenance genes was observed between GL and OE5. Conversely, four out of five genes of DNA demethylation showed significantly different expression in GL and OE5 (Figure 7). Interestingly, the expression of ROS1C/DNG701 and ROS1D/DNG704 was suppressed, whereas that of ROS1A/DNG702 was increased. These three genes were recently reported to demethylate DNA in the gamete and zygote, which is crucial for zygote gene expression and development (Zhou et al., 2021b). In addition, a mutation in ROS1A/DNG702 can generally lead to the increase of CG and CHG but not of CHH hypermethylation on genomes of rice endosperms (Liu J. et al., 2018). This result strongly corroborates our data wherein DMRs were only identified in CG and CHG sequence contexts, and because the expression of ROS1A/DNG702 was increased, markedly more hypo-DMRs (55.1%) and upregulated DEGs (61.5%) were observed (Figure 7). The effect of ROS1A/DNG702 may be counteracted by the decreased expression of ROS1C/DNG701 and ROS1D/DNG704. Taken together, we believe that ROS1A/DNG702 was the upstream protein that demethylated the promoter regions of OsHXX1 and
enhanced its expression, thus leading to decreased starch content and reduced rice growth.

DISCUSSION

Circadian Rhythm of OsPRR37 Is Important for Rice Growth
The endogenous expression period of clock genes is crucial for a plant to match the light–dark cycle. If correctly matched, the plant circadian system will enhance photosynthetic carbon fixation and growth (Dodd et al., 2005). Our results found that the circadian expression pattern of OsPRR37 in OE5 and OE9 was significantly different from that in GL and NIL-OsPRR37 (Figure 1). Although GL contains a loss-of-function allele of OsPRR37, the circadian rhythm and plant growth observed were similar between GL and NIL-OsPRR37 (Figure 1). Moreover, GL and NIL-OsPRR37 showed no significant difference in the starch content (Figure 6C). These results confirmed that disturbing circadian rhythm of OsPRR37 decreased starch content and plant growth.

Input and Output Pathways for OsPRR37
The regulatory network of transcription-translation feedback loops in the core circadian clock is well drawn based on exciting results of research on clock genes (Nakamichi, 2020). However, the inputs and outputs of the circadian clock remain unclear. The photosynthetic endogenous sugar levels provide metabolic entrainment to the circadian clock system through the morning-phased gene PRR7, the homolog of OsPRR37 in Arabidopsis (Haydon et al., 2013). A recent study reported that PRR7 mediates the circadian input to the promoter of CCA1 in
the shoots (Nimmo and Laird, 2021). These results indicate an
entrainment route of sugar–PRR7–CCA1. In rice, sugars suppress
OsCCA1 expression while OsCCA1 regulates IPA1 expression
to mediate panicule and grain development (Wang et al., 2020).
Our results suggested that the SDT/OsmiR156h–IPA1 module was
involved in modulating OsPRR37-mediated rice growth. Based
on these results, whether and how sugar–OsPRR37–OsCCA1–
SDT/OsmiR156h–IPA1 comprises an integrated pathway need
more evidence in the future. Furthermore, the enrichment
analysis found that upregulated genes were enriched in
carbohydrate metabolic process (Figure 4C), amino sugar and
nucleotide sugar metabolism, and carbon metabolism pathways
(Figure 4E). These results suggested that sugar and carbon
metabolism pathways are altered by OsPRR37 overexpression.
Meanwhile, several downregulated genes are enriched in nitrate
assimilation (Figure 4D) and nitrogen metabolism (Figure 4F),
suggesting that nitrate assimilation and metabolism would be
other pathways coordinated by OsPRR37 to affect plant growth.

The Role of Differentially Methylated
OsPRR37-Output Genes

Epigenetic modifications are closely associated with alterations
in chromatin structure, such as histone modification and DNA
methylation. Rhythmic transcription of Arabidopsis clock genes
was considered to be regulated by rhythmic histone modification
(Song and Noh, 2012). However, to our knowledge, there
has been no research on the role of DNA methylation in
regulating clock output genes. OsPRR37 was believed to repress
morning-phased output genes and indirectly activate evening-
phase output genes (Liu C. et al., 2018). In the present study,
as we focused on samples in the morning (9:00), and our
primary goal was to identify the key overlapping genes that
were downregulated by OsPRR37. In this process, we hoped to
get some insight into how OsPRR37 is associated with DNA
methylation pathways so as to directly repress the morning-
phased output genes. However, the results showed that 25 out
of 35 overlapping genes were upregulated, and the expression
levels of 22 genes were negatively correlated with methylation
levels (Figures 5B,C). These results supported our hypothesis
that DNA methylation contributed to the regulation of OsPRR37-
output genes, but the dynamic methylation of these output genes
is probably under an indirect regulation of OsPRR37. In other
words, the differentially methylated output genes are in the most
downstream of OsPRR37, such as OsHXK1, SDT/Osmir156h,
and OsMADS18, which are more directly to regulate rice growth,
flowering, and yield.

The Hierarchical Regulation Network
of OsPRR37

Different members of PRRs are supposed to function at their
specific times of the day to repress clock output genes (Farre
and Liu, 2013). Accumulating evidence has indicated that PRRs
interact with other proteins to regulate the transcription of output
genes. The B-box (BBX)-containing proteins BBX19 and BBX18
can physically interact with PRR9, PRR7, and PRR5 in a precise
temporal order from dawn to dusk, thus cooperatively regulating
the output genes (Yuan et al., 2021). OsPRR73 interacts with
histone deacetylase 10 (HDAC10) to co-repress OshKT21, a
plasma membrane-localized Na(+) transporter, and confers salt
stress tolerance to rice (Wei et al., 2020). The OsPRR37 protein
can interact with Ghdh8 and NF-YCs, which form an alternative
OsNF-Y heterotrimer to affect Hd1-mediated regulation of Hd3a
and flowering (Goretti et al., 2017). The distinct role of OsPRR37
in the ZH11 background indicated that OsPRR37 can associate
with different partners to perform different functions (Hu et al.,
2021). Moreover, the 35 identified differentially methylated DEGs
accounted for only a small proportion of DEGs (Figure 5A).
These results draw a map of the hierarchical regulation network
for OsPRR37 and thus put forward an interesting question about
the partners of OsPRR37 with which it regulates the large amount
of remaining DEGs. Nevertheless, differentially methylated DEGs
are the key candidates to regulate rice growth.

Epigenetic marks that modulate the expression of genes
behind the traits of interest have potential applications in crop
enhancement (Kakoulidou et al., 2021). With the development of
multi-omics technologies and related data processing pipelines
(Feng et al., 2021; Iqbal et al., 2021), the hierarchical regulation
network of the circadian clock will be gradually parsed and
applied to improve rice traits. Recently, the representative
role of OsPRR37 in the control of photoperiodic flowering
was systematically reviewed (Chen et al., 2021; Zhou et al.,
2021c). However, the underlying mechanism of how OsPRR37
regulates its output genes to affect multiple agronomic traits
remains unclear. By integrative analysis of WGBS and RNA-
seq data, our results revealed that DNA methylation contributes
to the regulation of OsPRR37-output genes, which provides an
alternative strategy to improve plant growth through epigenetic
modulation of OsPRR37-output genes.

DATA AVAILABILITY STATEMENT

The datasets presented in this study can be found in online
repositories. The name of the repository and accession number
can be found below: GEO, NCBI: GSE192416.

AUTHOR CONTRIBUTIONS

CL designed the study, carried out most of the data analysis,
and wrote the manuscript. NL performed the qRT-PCR and
determined the starch content. ZL, QS, XP, XX, CD, and
ZX performed the data analysis of time-course transcriptomes
and tissue-specific microarray data. KS, FY, and ZH provided
insightful suggestions on data analysis and writing of manuscript.
All authors have read and agreed to the submitted version
of the manuscript.

FUNDING

This work was funded by Chongqing Postdoctoral Science
Foundation (2018LY10), Chongqing Natural Science Foundation
ACKNOWLEDGMENTS

We wish to thank Daichang Yang of the Wuhan University for providing the field area to plant transgenic rice. We also wish to thank Jianming Zeng of the University of Macau and his biotrainee team for providing bioinformatics courses.

SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fpls.2022.839457/full#supplementary-material
DNA Methylation Regulates OsPRR37-Output Genes

Vera Alvarez, R., Pongor, L. S., Marino-Ramirez, L., and Landsman, D. (2019). TPMCalculator: one-step software to quantify mRNA abundance of genomic features. Bioinformatics 35, 1960–1962. doi: 10.1093/bioinformatics/bty196

Wang, F., Han, T., Song, Q., Ye, W., Song, X., Chu, J., et al. (2020). The rice circadian clock regulates tiller growth and panicle development through strigliolate signaling and sugar sensing. Plant Cell 32, 3124–3138. doi: 10.1105/tpc.20.00289

Wang, H., Li, Y., Chern, M., Zhu, Y., Zhang, L. L., Lu, J., et al. (2021). Suppression of rice miR168 improves yield, flowering time and immunity. Nat. Plants 7, 129–136. doi: 10.1038/s41477-021-00885-x

Wang, L., Xie, W., Chen, Y., Tang, W., Yang, J., Ye, R., et al. (2010). A dynamic gene expression atlas covering the entire life cycle of rice. Plant J. 61, 752–766. doi: 10.1111/j.1365-313X.2009.04100.x

Wei, H., Wang, X., He, Y., Xu, H., and Luan, L. (2020). Clock component OsPRR37 positively regulates rice salt tolerance by modulating OsHKT2;1-mediated sodium homeostasis. EMBO J. 40, e105086. doi: 10.15252/embj.2020105086

Wu, J., Yang, Z., Wang, Y., Zheng, L., Ye, R., Ji, Y., et al. (2015). Viral-inducible Argonauta18 confers broad-spectrum virus resistance in rice by sequestering a host microRNA. Elife 4:e05733. doi: 10.7554/eLife.05733

Wu, T., Hu, E., Xu, S., Chen, M., Guo, P., Dai, Z., et al. (2021). clusterProfiler 4.0: a universal enrichment tool for interpreting omics data. Innovation 2:100141. doi: 10.1016/j.innov.2021.100141

Xu, L., Yuan, K., Yuan, M., Meng, X., Chen, M., and Luan, J. (2020). Regulation of rice tillering by RNA-directed DNA methylation at inverted repeat-transposable elements. Mol. Plant 13, 851–863. doi: 10.1016/j.molp.2020.02.009

Yan, W., Liu, H., Zhou, X., Li, Q., Zhang, J., Lu, L., et al. (2013). Natural variation in GhD7:1 plays an important role in grain yield and adaptation in rice. Cell Res. 23, 969–971. doi: 10.1038/cr.2013.43

Yang, M., Li, Y., Liu, Z., Tian, J., Liang, L., Qu, Y., et al. (2020). A high activity zinc transporter OsZIP9 mediates zinc uptake in rice. Plant J. 103, 1695–1709. doi: 10.1111/tpj.14855

Yang, S. Y., Gronlund, M., Jakobsen, L., Grottemeyer, M. S., Rentsch, D., Miyao, A., et al. (2012). Nonredundant regulation of rice arbuscular mycorrhizal symbiosis by two members of the phosphate transporter gene family. Plant Cell 24, 4236–4251. doi: 10.1105/tpc.112.104901

Yao, W., Li, G., Yu, Y., and Ouyang, Y. (2018). funRiceGenes dataset for comprehensive understanding and application of rice functional genes. Gigaservice 7, 1–9. doi: 10.1093/gigaservice/gix119

Yin, X., Liu, X., Xu, B., Lu, P., Dong, T., Yang, D., et al. (2019). OsMADS51, a membrane-bound MADS-box transcription factor, modulates plant architecture and the abscisic acid response in rice. J. Exp. Bot. 70, 3895–3909. doi: 10.1093/jxb/erz198

Yi, G., Wang, L. G., and He, Q. Y. (2015). ChlPeeker: an R/Bioconductor package for ChlP peak annotation, comparison and visualization. Bioinformatics 31, 2382–2383. doi: 10.1093/bioinformatics/btv415

Yuan, L., Yu, Y., Liu, M., Song, Y., Li, H., Sun, J., et al. (2021). BBX19 fine-tunes the circadian rhythm by interacting with PSEUDO-RESPONSE REGULATOR proteins to facilitate their repressive effect on morning-phased clock genes. Plant Cell 33, 2602–2617. doi: 10.1105/tpc.2020.1105/tpc.20.00289

Zeng, D. D., Yang, C. C., Qin, R., Alamin, M., Yue, E. K., Jin, X. L., et al. (2018). A guanine insert in OsBS51 leads to early leaf senescence and salt sensitivity in rice (Oryza sativa L.). Plant Cell Rep. 37, 933–946. doi: 10.1007/s00207-018-2280-y

Zhang, C., Wei, Y., Xu, L., Wu, K. C., Yang, L., Shi, C. N., et al. (2020). A Bunyaviruses-inducible ubiquitous ligase targets RNA polymerase IV for degradation during viral pathogenesis in rice. Mol. Plant 13, 836–850. doi: 10.1016/j.molp.2020.02.010

Zhang, H., Lang, Z., and Zhu, J. K. (2018). Dynamics and function of DNA methylation in plants. Nat. Rev. Mol. Cell Biol. 19, 489–506. doi: 10.1038/s41580-018-0016-z

Zhang, X., Sun, J., Cao, X., and Song, X. (2015). Epigenetic mutation of RAV6 affects leaf angle and seed size in rice. Plant Physiol. 169, 2118–2128. doi: 10.1104/pp.150.038586

Zhang, X., Yazaki, J., Sundaresan, A., Cokus, S., Chan, S. W., Chen, H., et al. (2006). Genome-wide high-resolution mapping and functional analysis of DNA methylation in Arabidopsis. Cell 126, 1189–1201. doi: 10.1016/j.cell.2006.08.003
Zheng, S., Li, J., Ma, L., Wang, H., Zhou, H., Ni, E., et al. (2019). OsAGO2 controls ROS production and the initiation of tapetal PCD by epigenetically regulating OsHXK1 expression in rice anthers. *Proc. Natl. Acad. Sci. U. S. A.* 116, 7549–7558. doi: 10.1073/pnas.1817675116

Zheng, S., Ye, C., Lu, J., Liufu, J., Lin, L., Dong, Z., et al. (2021). Improving the rice photosynthetic efficiency and yield by editing OsHXK1 via CRISPR/Cas9 system. *Int. J. Mol. Sci.* 22:9554. doi: 10.3390/ijms22179554

Zhou, J., Zhang, R., Jia, X., Tang, X., Guo, Y., Yang, H., et al. (2021a). CRISPR-Cas9 mediated OsMIR168a knockout reveals its pleiotropy in rice. *Plant Biotechnol. J.* 20, 310–322. doi: 10.1111/pbi.13713

Zhou, S., Li, X., Liu, Q., Zhao, Y., Jiang, W., Wu, A., et al. (2021b). DNA demethylases remodel DNA methylation in rice gametes and zygote and are required for reproduction. *Mol. Plant* 14, 1569–1583. doi: 10.1016/j.molp.2021.06.006

Zhou, S., Zhu, S., Cui, S., Hou, H., Wu, H., Hao, B., et al. (2021c). Transcriptional and post-transcriptional regulation of heading date in rice. *New Phytol.* 230, 943–956. doi: 10.1111/nph.17158

Zhu, Q., Ordiz, M. I., Dabi, T., Beachy, R. N., and Lamb, C. (2002). Rice TATA binding protein interacts functionally with transcription factor II1B and the RF2a bZIP transcriptional activator in an enhanced plant in vitro transcription system. *Plant Cell.* 14, 795–803. doi: 10.1105/tpc.010364

**Conflict of Interest:** The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

**Publisher’s Note:** All claims expressed in this article are solely those of the authors and do not necessarily represent those of their affiliated organizations, or those of the publisher, the editors and the reviewers. Any product that may be evaluated in this article, or claim that may be made by its manufacturer, is not guaranteed or endorsed by the publisher.