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Special Issue Article

Does biological rhythm transmit from plants to rhizosphere microbes?

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Summary

Plant physiological and metabolic processes are modulated by rhythmic gene expression in a large part. Meanwhile, plants are also regulated by rhizosphere microorganisms, which are fed by root exudates and provide beneficial functions to their plant host. Whether the biorhythms in plants would transfer to the rhizosphere microbial community is still uncertain and their intricate connection remains poorly understood. Here, we investigated the role of the *Arabidopsis* circadian clock in shaping the rhizosphere microbial community using wild-type plants and clock mutants (*cca1-1* and *toc1-101*) with transcriptomic, metabolomic and 16S rRNA gene sequencing analysis throughout a 24-h period. Deficiencies of the central circadian clock led to abnormal diurnal

rhythms for thousands of expressed genes and dozens of root exudates. The bacterial community failed to follow obvious patterns in the 24-h period, and there was lack of coordination with plant growth in the clock mutants. Our results suggest that the robust rhythmicity of genes and root exudation due to circadian clock in plants is an important driving force for the positive succession of rhizosphere communities, which will feedback on plant development.

Introduction

The endogenous circadian clock in plants regulates most of the transcriptome and provides temporal coordination to ensure the optimal efficiency of many biological functions, including photosynthesis, carbohydrate metabolism and reproduction (Harmer *et al.*, 2000; Michael *et al.*, 2008; Carbonell-Bejerano *et al.*, 2014). Plant rhizosphere microbes play a role in plant growth by regulating soil fertility, plant health and productivity (Peiffer *et al.*, 2013; Wagg *et al.*, 2014; Muller *et al.*, 2016; Castrillo *et al.*, 2017; Lu *et al.*, 2018). The composition of the rhizosphere microbial community is predominantly determined by soil properties and environmental factors, but it is also strongly affected by the genotype of the plant host (Lundberg *et al.*, 2012; Chaparro *et al.*, 2014; Ofek-Lalzar *et al.*, 2014; Qu *et al.*, 2020).

The current concept of plant–microbiome holobiont considers the multicellular host and its associated microbiota as a functional entity (Rosenberg and Zilber-Rosenberg, 2016; Hassani *et al.*, 2018). We can safely infer that the diurnal rhythmicity of the plants and microbes within the holobiont would influence each other, regulating plant growth and the composition of the microbial community. The structure of the bacterial community in the rhizosphere of *Arabidopsis* differs substantially between samples collected during the light and dark cycle, and this diurnal variation has been associated with the metabolism and exchange of carbon in the plants and microbes (Staley *et al.*, 2017; Hubbard *et al.*, 2018). The molecular mechanisms involving a strong

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relationship between changes to the microbial community and the plant circadian clock nevertheless remain poorly understood. Understanding the form and extent to which plant diurnal rhythms influence microorganisms has thus become necessary.

The circadian clock in *Arabidopsis* involves a post-transcriptional component and a well-studied transcriptional–translational system of feedback regulation, in which circadian clock associated 1 (CCA1) and timing of CAB2 expression 1 (TOC1) are two core genes (Alabadi *et al.*, 2001; Huang *et al.*, 2012; Pokhilko *et al.*, 2012). Mutants of these two genes disrupt the diurnal rhythm (Gendron *et al.*, 2012; Nagel *et al.*, 2015). We cultivated *cca1-1* and *toc1-101* mutants with the wild type (Col-0) under the same conditions to gain a better understanding of the role that the diurnal rhythm of plants plays in the control of the rhizosphere microbial community. We performed transcriptomic and metabolomic analyses and 16S rRNA gene sequencing to comprehensively investigate whether the diurnal rhythm affected the composition of the microbial community and, if so, whether this effect was intimately linked with root exudates.

Materials and methods

Experimental design and seedling culture

We used wild type (Col-0) and *cca1-1* and *toc1-101* mutants (both deficient in central transcriptional–translational feedback loops of the circadian clock) of *Arabidopsis thaliana* to determine the role of plant circadian clock in determining the nature of the rhizosphere microbiome. The surfaces of *Arabidopsis* seeds were sterilized to prevent bacterial contamination of solid media and were vernalized after that. The seeds were then cultured in Petri dishes containing Murashige and Skoog (MS) medium under sterile conditions at 25°C at a light intensity of 300 $\mu\text{mol photons m}^{-2} \text{ s}^{-1}$ and a 12:12-h light:dark photoperiod. The MS medium contained 3% sugar and 0.5% agar and was autoclaved at 115°C for 30 min before use. Two-week-old aseptic seedlings were transplanted into polycarbonate pots (400 ml) containing autoclaved potting soil (Sun Gro Horticulture, Avagam, MA, USA) and soil slurry. Approximately 30 g of grassland soil collected near the Zhejiang University of Technology, China (30°17'45.11"N, 120°09'50.07"E) was mixed with 200 ml of sterile water by vigorous shaking for 60 s. Eighteen millilitres of soil slurry were added to each pot before transplantation. Plants were grown in an artificial greenhouse at 25 \pm 0.5°C and 80% relative humidity under cool-white fluorescent light (300 $\mu\text{mol photons m}^{-2} \text{ s}^{-1}$) with a 12:12-h light:dark cycle. Samples for transcriptomic and metabolomic analyses and 16S rRNA gene sequencing

were collected 14 days after transplantation. The timepoints for the analysis of diurnal rhythmicity were 07:00 [1 h before the start of the light exposure, Zeitgeber time (ZT) 0], 13:00 (ZT6), 19:00 (1 h before the dark cycle, ZT12), 01:00 (next day, ZT18) and 07:00 (next day, ZT24). For each genotype at each timepoint, 15 pots were used. The details of the sample collection are provided below.

Sampling of plant shoots and RNA isolation

About 0.1 g of aboveground plant tissue was weighed at each timepoint and placed into a 1.5-ml centrifuge tube, frozen in liquid nitrogen and stored at -80°C . Three replicate samples were collected for each genotype for transcriptomic analysis.

TRIzol[®] RNA reagent was used to extract total RNA from the samples following the manufacturer's instructions (Invitrogen, Carlsbad, CA, USA). Genomic DNA was removed using DNase I (TaKaRa), RNA quality was assessed using a 2100 Bioanalyzer (Agilent) and RNA was quantified using an ND-2000 spectrophotometer (NanoDrop Technologies). Only samples with high-quality RNA (OD260/280 = 1.8–2.2, OD260/230 \geq 2.0, RIN \geq 6.5, 28S:18S \geq 1.0, total RNA > 2 μg) were used to construct a sequencing library.

Library preparation and RNA sequencing

RNA purification, reverse transcription, library construction and sequencing were all performed at Shanghai Majorbio Bio-pharm Biotechnology (Shanghai, China) following the manufacturer's instructions (Illumina, San Diego, USA). One microgram of total RNA for each genotype was used to prepare the RNA-seq transcriptomic library using the TruSeq[™] RNA kit (Illumina) for sample preparation. Messenger RNA was isolated by using the polyA-selection method and oligo(dT) beads and then fragmented in a fragmentation buffer. A SuperScript double-stranded cDNA synthesis kit (Invitrogen) with random hexamer primers (Illumina) was used to synthesize double-stranded cDNA. The synthesized cDNA was end-repaired and phosphorylated, and adenines were added to the 3' end following Illumina's protocol for constructing libraries. Libraries were size selected for cDNA target fragments of 200–300 bp on 2% Low Range Ultra Agarose followed by PCR amplification using Phusion DNA polymerase (New England Biolabs, Boston, MA, USA) for 15 cycles with the default parameters mentioned in the instruction of Phusion DNA polymerase. A paired-end RNA-seq library was sequenced using the Illumina Novaseq 6000 system (2 \times 150 bp read length) after quantification using a TBS380 mini-fluorometer.

Analysis of the RNA-seq data

The sequences were trimmed and quality controlled using SeqPrep (<https://github.com/jstjohn/SeqPrep>) and Sickie (<https://github.com/najoshi/sickle>) using default parameters, and clean reads were separately aligned to the *A. thaliana* reference genome using TopHat (<http://tophat.cbcb.umd.edu/>; version 2.1.1) in orientation mode. We gained more than 7 Gb clean data per sample and only reads uniquely matched to the genome, allowing ≤ 2 mismatches without insertions or deletions, were mapped. The numbers of fragments per kilobase of exon per million mapped reads (FPKM) were calculated to represent the expression level. The ARSER algorithm (Yang and Su, 2010) was used to detect diurnal rhythmic genes using the parameters: minimal period, 20; maximal period, 28; default period, 24 and $P = 0.05$. A functional-enrichment analysis of the selected gene sets was performed to detect significantly enriched genes in the Kyoto Encyclopedia of Genes and Genomes (KEGG) relative to the whole-transcriptome background. The KEGG pathway analysis was carried out using KOBAS 2.1.1 (<http://kobas.cbi.pku.edu.cn/download.php>).

Preparation of exudate samples

Six replicates were taken from each genotype, and six seedlings were taken from each parallel. The plant was removed by cutting the polycarbonate pots which *Arabidopsis* planted in and its rhizosphere was gently non-destructively removed by shaking. The remaining soil attached to the roots was removed by washing with ddH₂O, and the roots were then placed in a prepared petri dish (with an orifice plate) containing 40 ml of sterile water. Each root was allowed to secrete for 6 h each replicate. The secretions were then collected in a 50-ml centrifugal tube using a 0.45- μ m aqueous filter head, quickly frozen in liquid nitrogen and stored at -80°C . The weight of the plant and the volume of the secretions were recorded. The samples were thawed on ice, and the exudates were extracted in a 50% methanol buffer. The exudates were centrifuged at 4000g for 20 min, and the supernatants were transferred to new 96-well plates. Samples for quality control (QC) were prepared by pooling equal volumes (10 μ l) from all samples. The samples were stored at -80°C prior to analysis by liquid chromatography and mass spectroscopy (LC–MS).

Measurement and identification of root exudates

All chromatographic separations were performed using an ultraperformance liquid chromatograph (UPLC) (SCIEX, UK). An ACQUITY UPLC T3 column (100 mm \times 2.1 mm, 1.8 μ m; Waters, UK) was used for

the reversed-phase separation. The column oven was maintained at 35°C , and the flow rate was 0.4 ml min⁻¹. The mobile phase consisted of solvents A (0.1% formic acid) and B (acetonitrile, 0.1% formic acid) with the following conditions for gradient elution: 0–0.5 min, 5% B; 0.5–7 min, 5%–100% B; 7–8 min, 100% B; 8–8.1 min, 100%–5% B; and 8.1–10 min, 5% B. Exudates eluted from the column were detected using a TripleTOF5600plus high-resolution tandem mass spectrometer (SCIEX, Boston, MA, USA) in both positive- and negative-ion modes. The ionspray voltage floating was set at 5000 and -4500 V for the positive- and negative-ion modes, respectively. The MS data were acquired in IDA mode, and the TOF mass ranged from 60 to 1200 Da. The survey scans were acquired in 150 ms, and as many as 12 product ion scans were collected if exceeding a threshold of 100 counts second⁻¹ and with a +1 charge. Total cycle time was fixed to 0.56 s. Four-time bins were summed for each scan at a pulse frequency of 11 kHz by monitoring the 40-GHz multichannel TDC detector with four-anode/channel detection. Dynamic exclusion was set for 4 s. Mass accuracy was calibrated during the acquisition every 20 samples. A QC sample was collected after every 10 samples to evaluate the stability of the LC–MS throughout the acquisition. Raw LC–MS data files were converted to mzXML format and processed using the XCMS, CAMERA and metaX toolboxes in R. Each ion was identified by combining the retention times and m/z data. Exudates with mass differences between the observed and database values of <10 ppm were annotated using the online KEGG and HMDB databases. The molecular formulae of the exudates were determined and validated by measurements of isotopic distribution.

Analysis of metabolomic data

The intensity of the peaks of the exudates was calculated by the area of the peaks and preprocessed using metaX, and peaks were removed when detected in <50% of the QC samples or <80% of the biological samples. The remaining peaks with missing values were calculated with the k-nearest neighbour algorithm to further improve the quality of the data. The preprocessed data set was used for detecting outliers and evaluating batch effects using a principal component analysis (PCA). A robust QC-based LOESS signal correction was fitted to the QC data in the order of injection to minimize the drift of signal intensity over time. Metabolic features with relative standard deviations >30% across all QC samples were removed. A principal coordinate analysis (PCoA) combined with a permutational analysis of variance (PERMANOVA) were performed to evaluate the dissimilarity between genotypes using Bray–Curtis distances.

The ARSER algorithm was used to detect diurnal rhythmic metabolites using the parameters: minimal period, 20; maximal period, 28; default period, 24 and $P = 0.05$. Spearman's rank correlations (r) between genus were calculated using the psych package in R. Only robust ($r > 0.8$ or $r < -0.8$) and statistically significant (P -value < 0.05) correlations were incorporated into co-occurrence network. Network visualization and modular analysis were made with Gephi (v 0.9.2). Node-level topological properties were also calculated using Gephi.

Collection of rhizospheres and bulk soils

The soil around the roots was gently removed to minimize mechanical damage to the roots. The roots were then placed in a 50-ml centrifuge tube filled with 20 ml of PBS buffer (137 mM NaCl, 10 mM phosphate buffer, 2.7 mM KCl, pH 7.3–7.5). The tube was then placed on a shaker at 180 rpm for 20 min, and the roots were discarded using sterile tweezers. The washing solution was centrifuged at 3900g for 20 min. The supernatant was discarded, retaining the rhizosphere (Bulgarelli *et al.*, 2012). The bulk soil (soil without roots) from the same pot as each sample of rhizosphere was also collected. Samples were stored at -80°C prior to 16S rRNA gene sequencing. Four parallel samples were collected for each genotype for sequencing.

16S rRNA gene sequencing

A FastDNA SPIN Kit for soil (MP Biomedicals, CA, USA) was used to extract total DNA from the microorganisms in the rhizosphere and bulk soil following the manufacturer's instructions. The V3–V4 region of the 16S rRNA gene was amplified using the primers 338F (5'-ACTCCTACGGGAGGCAGCAG-3') and 806R (5'-GGACTACHVGGGTWTCTAAT-3') and a PCR thermocycler (GeneAmp 9700; ABI, USA). Purified amplicons were sequenced on an Illumina MiSeq platform (Illumina). Fastp v0.14.1 (<https://github.com/OpenGene/fastp>) was used to confirm the quality of raw data, the sequences with lower quality and the primers were removed by using Trimmomatic v.0.36 (<http://www.usadellab.org/cms/index.php?page=trimmoMatic>) and Cutadapt v2.10 (<https://github.com/marcelm/cutadapt/>), respectively. Paired-end clean reads were merged by using Usearch v10 (<http://www.drive5.com/usearch/>). After subsampling each sample to an equal sequencing depth (22 928 reads per sample), sequences with $\geq 97\%$ similarity were assigned to the same operational taxonomic unit (OTU). In total, we gained 5970 OTUs in this study. Taxa were assigned using the SILVA database (<https://www.arb-silva.de/>) based on the RDP classifier (Version 2.11, <https://sourceforge.net/projects/rdp-classifier/>) algorithm. The

microbial compositions were also assessed using the UniFrac distance, which is a phylogenetically based measure of the degree of similarity between microbial communities (Lozupone and Knight, 2005). Pairwise distances were determined for all pairs of samples based on either the taxonomic representation (unweighted) or the types and relative abundances (weighted). A PCoA of the UniFrac distances was used to visualize the differences among samples. A PERMANOVA was performed to evaluate the variance between genotypes. The ARSER algorithm was used to detect genera with diurnal rhythms using the parameters: minimal period, 20; maximal period, 28; default period, 24 and $P = 0.05$.

Analysis of transcriptomes, root exudates and 16S rRNA gene sequences

The Procrustes test was used to analyse the correlations between rhythmic genes in the wild type and all exudates using the vegan package in R (3.6.2). Multiple linear regression was used to analyse the regulation of all exudates from the 10 most abundant phyla (Proteobacteria were divided into three classes) using SPSS (v 20.0.0).

Results

Physiological differences between the wild-type Arabidopsis and clock mutants

To explore differences in plant growth caused by an abnormal circadian clock, we measured the fresh weights and bolting rate of the wild type (Col-0) and the two mutants (*cca1-1* and *toc1-101*). The fresh weights of 28-day-old seedlings of the wild type and the two mutants were similar and did not differ significantly (Supporting Information Fig. S1). The bolting rate during 15–28 days of culture was slightly lower (5%–10%) for the *toc1-101* mutant than the wild type, and the bolting rate was much higher (22%–68%) for the *cca1-1* mutant than the two other lines (Supporting Information Fig. S2).

Diurnal transcriptomic variation between wild-type Arabidopsis and clock mutants

To further investigate the detailed cycling dynamics of transcription, we sampled leaves every 6 h over a 24-h period (from 07:00 to next 07:00), for a total of five timepoints, hereafter designated as ZT 0–24, from the wild type and mutants for a transcriptomic analysis (Fig. 1A). The diurnal rhythmicity of the genes in both the wild type and mutants was identified using the ARSER algorithm (Yang and Su, 2010; Wu *et al.*, 2016) (Fig. 1B). Heatmaps of the patterns of cycling genes (ARSER, $P < 0.05$) clearly differed between the two mutants and

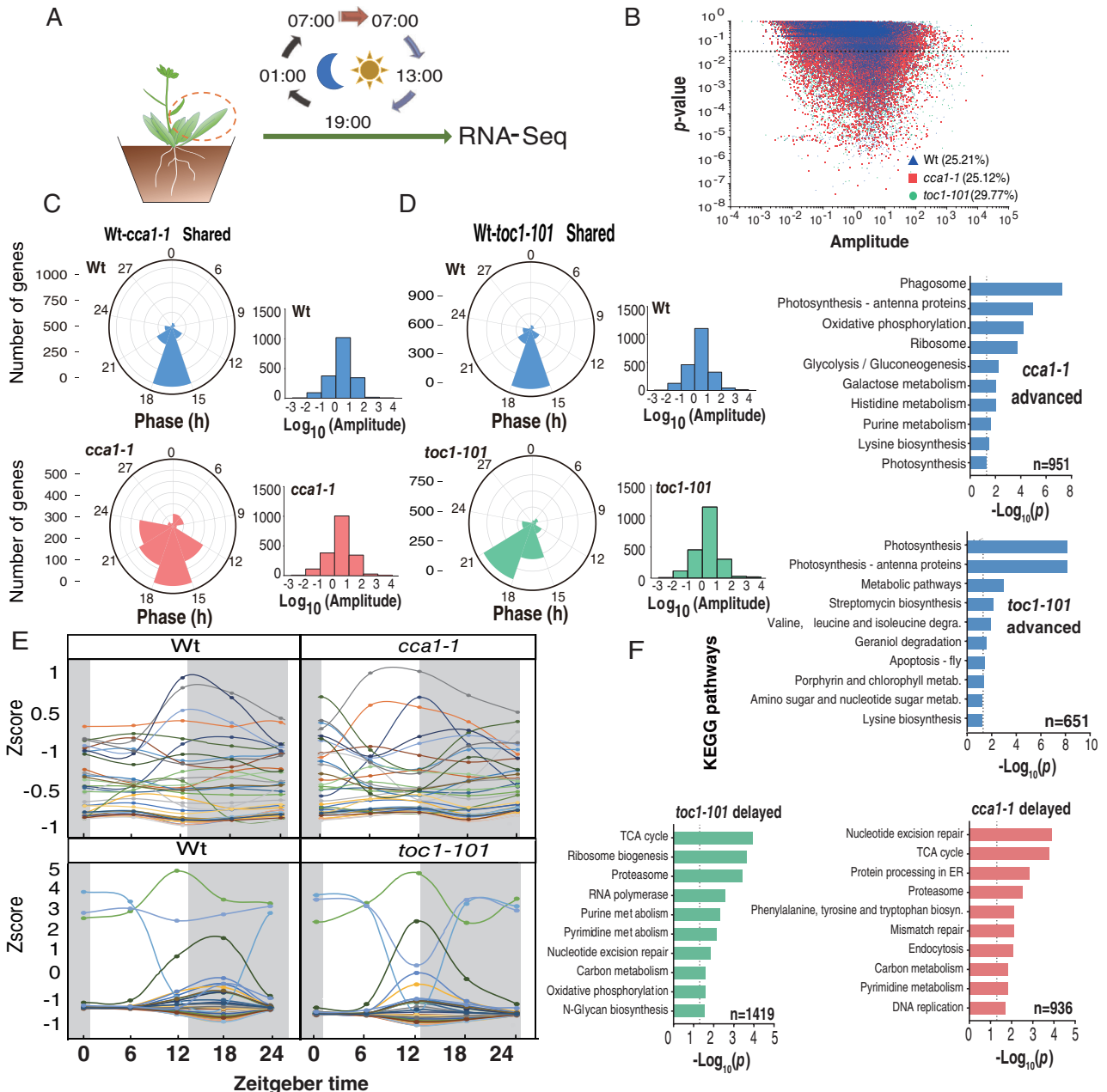


Fig. 1. Transcriptional oscillations in the wild type and the *cca1-1* and *toc1-101* mutants.

A. Schematic diagram of transcriptional oscillations.

B. Diurnal oscillations of all genes of the three genotypes, with each dot representing a gene and the dashed lines representing rhythmicity (ARSER, $P < 0.05$).

C. Distribution of the oscillating phase and amplitude of cycling genes shared between wild-type (*Wt*) *Arabidopsis* and the *cca1-1* mutant.

D. Distribution of the oscillating phase and amplitude of cycling genes shared between wild-type *Arabidopsis* and the *toc1-101* mutant.

E. Rhythmic oscillation in 24 h of cycling genes shared between wild-type *Arabidopsis* and the *cca1-1* (upper panel) and *toc1-101* (lower panel) mutants; note that only 50 genes were chosen based on the p value of the ARSER algorithm for the wild type.

F. KEGG analysis of cycling genes shared between wild-type *Arabidopsis* and the *cca1-1* or *toc1-101* mutant but with abnormal phases.

the wild type (Supporting Information Fig. S3A and C). The *cca1-1* mutant lost 4902 cycling genes but gained 4940 cycling genes compared to the wild type, similar to the *toc1-101* mutant. The detailed classification of the

lost and gained cycling genes in the *cca1-1* and *toc1-101* mutants (Supporting Information Fig. S3b and d) indicated that the rhythmic operation of multiple pathways, including many types of metabolic and transport

pathways, differed greatly from the wild type. The cycling genes shared between the wild type and the mutants were involved in photosynthesis, the TCA cycle, oxidative phosphorylation and nucleotide excision repair, indicating that the mutants retained the rhythmicity of the most basic activities for plant life. The growth (represented by fresh weight) and phenotypes of the mutants thus did not differ significantly from those of the wild type (Supporting Information Fig. S1).

The wild type and the two mutants shared 1800 cycling genes, but their oscillating loci were generally distinct (Fig. 1C–E). The peak expression of most cycling genes in the wild type ranged between ZT15 and ZT18. The expression of cycling genes peaked equally between ZT12 and ZT24 in the *cca1-1* mutant, but more genes peaked between ZT18 and ZT21 in the *toc1-101* mutant. Many genes shifted their phase in the mutants, although the oscillating amplitude (maximum diurnal variation in gene expression) of these genes was not significantly affected in the mutants. These phase-changed genes are mainly inferred to basic activities for plant life, including metabolism of lipids, carbohydrates, energy, and amino acids (Fig. 1F).

The circadian clock in *Arabidopsis* plays an important role in regulating the biosynthesis of central and secondary metabolites (Fukushima *et al.*, 2009). A Procrustes analysis indicated that the variation of the cycling in metabolic pathways, and changes to the phase shift and amplitude of oscillating genes, affected the synthesis and transport of material in the plant, which may further affect the root exudates (Supporting Information Fig. S4).

Distinct root exudates between wild-type Arabidopsis and clock mutants

We collected the root exudates of the three *Arabidopsis* lines (wild type and *cca1-1* and *toc1-101* mutants) at five timepoints, with timepoint representing the continuous secretion of exudates for the previous 6 h, e.g., the exudates sampled at ZT12 consisted of the exudates secreted between ZT6 and ZT12 (Fig. 2A). A total of 196 exudates were identified (Supporting Information Data set S1), 42% of which were lipids and lipid-like molecules. The daily production (total of all timepoints) of some exudates differed between the wild type and the mutants. A global PCoA combined with an Adonis analysis (Fig. 2B) found that both genotype ($R^2 = 0.04$, $P < 0.01$) and timepoint ($R^2 = 0.33$, $P < 0.001$) could be linked to the profile of exudates in *Arabidopsis*. The exudates in the wild type differed significantly between ZT0 and ZT6 (Supporting Information Fig. S5). The differences between timepoints were not as obvious in the *cca1-1* mutant as the differences in the wild type, although the Adonis analysis identified a significant

difference between timepoints ($R^2 = 0.55$, $P < 0.001$). Exudates for the *toc1-101* mutant at ZT0, ZT6 and ZT12 differed from those secreted at other timepoints.

We detected 50 rhythmic exudates in the wild type using the ARSER algorithm (Figs. 2C and S6) but fewer in the *cca1-1* (39) and *toc1-101* (45) mutants. Some exudates retained rhythmicity in the mutants but had different rhythmic features (only four metabolites were concurrently rhythmic in all genotypes). The amplitude and phase for the cycling exudates shared between the mutants and the wild type varied with genotype (Fig. 2D).

The heatmap in Fig. S7A represents the number of cycling exudates shared between the *cca1-1* mutant and the wild type (10), with some exudates unique to the wild type (40) and the *cca1-1* mutant (29). The heatmap in the Supporting Information Fig. S7b represents the number of cycling exudates shared between the *toc1-101* mutant and the wild type (17), with some exudates unique to the wild type (33) and the *toc1-101* mutant (28). Exudates that varied their rhythmicity in the mutants were mainly lipids and lipid-like molecules. The five most rhythmic lipids and lipid-like molecules in the wild type were thus chosen based on their *P*-values calculated using the ARSER algorithm for demonstrating the effects of the central circadian clock on metabolism. The levels of these five exudates varied rhythmically in the wild type, but the exudates in the mutants lost their rhythmicity [dihydroterpineol, phosphatidylethanolamine (15:0/18:0) and piperochromenoic acid] or had abnormal amplitudes and phases (acylcarnitine 14:2 and citronellyl acetate) (Fig. 2E). Both the abnormal amplitude and phase of exudation might lead to different trends in development of the rhizosphere microbiome and would lead to distinct community structures in the three *Arabidopsis* lines.

Temporal changes of microbial community in rhizosphere and bulk soil

We sampled the rhizosphere soil of the wild type and the *cca1-1* and *toc1-101* mutants every 6 h for 24 h to assess the influence of the rhythmic behaviour of plants on the rhizosphere microbiome (Fig. 3A). An equal number of samples of rhizosphere and bulk soils were collected from the same pots, so each genotype at each timepoint had a bulk soil as a control. A PCoA based on unweighted UniFrac distances clearly indicated that the rhizosphere and bulk soils had distinct microbial communities (Fig. 3B). The initial soil microbial communities in the pots containing the cultivated *cca1-1* and *toc1-101* mutants and the wild type were from the same soil collected from a grassland. The composition of the microbiota in the bulk soil, however, was still very similar among the mutants and the wild type after 14 day of cultivation (Supporting Information Fig. S8A), but the

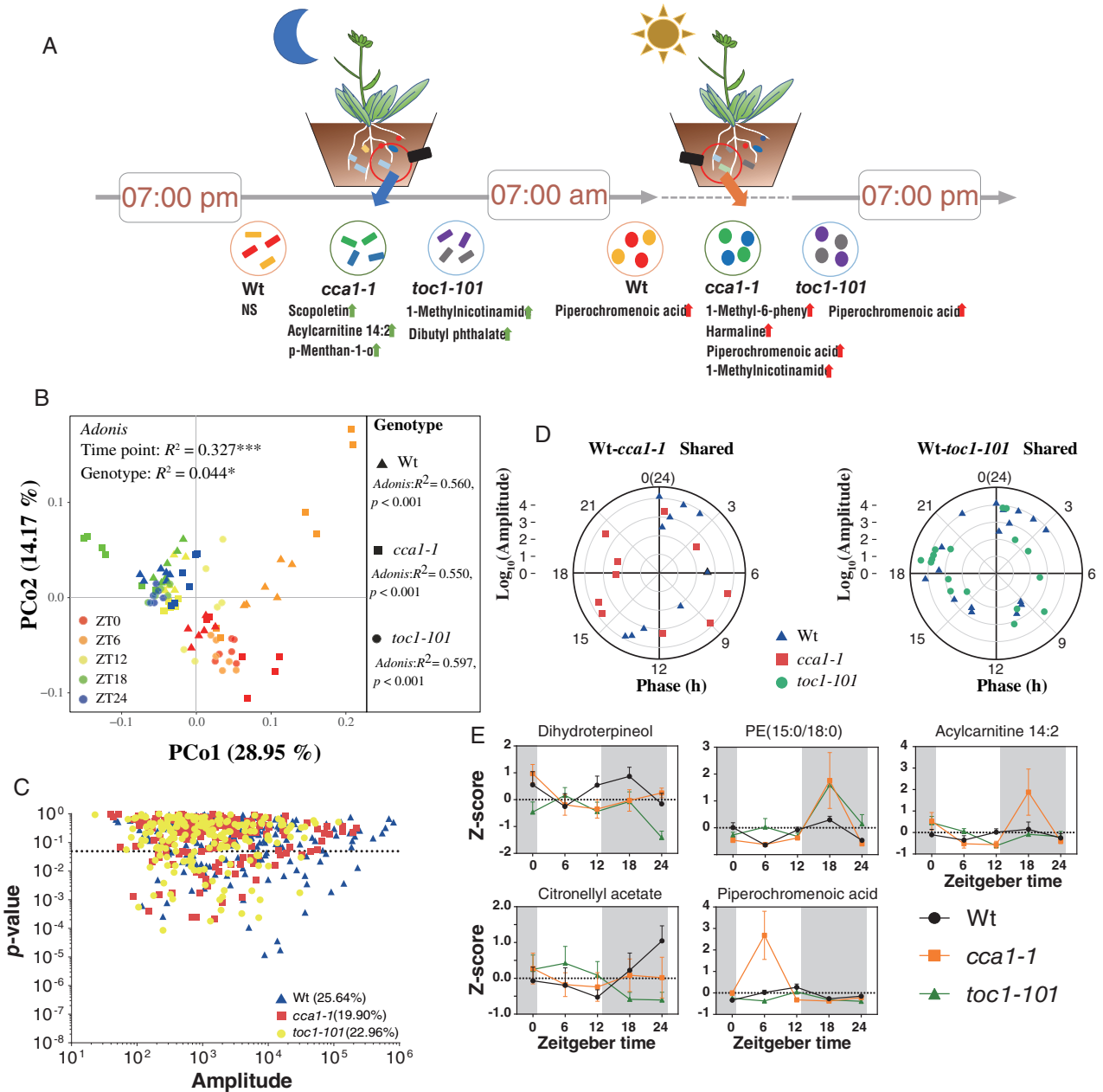


Fig. 2. Diurnal patterns of the root exudates.

A. Schematic of the diurnal patterns of the root exudates.

B. PCoA based on Bray–Curtis distances and a PERMANOVA evaluation of the dissimilarity among the three genotypes and five timepoints; the panel on the right shows the results of the PERMANOVA of the effect of time on the root exudates for each genotype.

C. Diurnal oscillations of root exudates for the three genotypes, with each dot representing an exudate and the dashed lines representing rhythmicity (ARSER, $P < 0.05$).

D. Distribution of the phase of cycling exudates shared between the wild type (Wt) and the *cca1-1* and *toc1-101* mutants.

E. Rhythmic oscillation of lipids and lipid-like molecules in 24 h; note that only five lipids and lipid-like molecules with the most significant rhythmicities were chosen based on the p value of the ARSER algorithm for the wild type.

microbial communities in the rhizosphere soils differed significantly between the mutants and the wild type (Supporting Information Fig. S8B).

The composition of the microbial community varied significantly over time (Adonis analysis with unweighted

UniFrac distance, $P = 0.002$) in the rhizosphere of the wild type but not the mutants (Supporting Information Fig. S9). The ARSER algorithm identified more rhythmic OTUs in the rhizosphere of the *cca1-1* (19.80%) and *toc1-101* (22.67%) mutants than the wild type (14.43%)

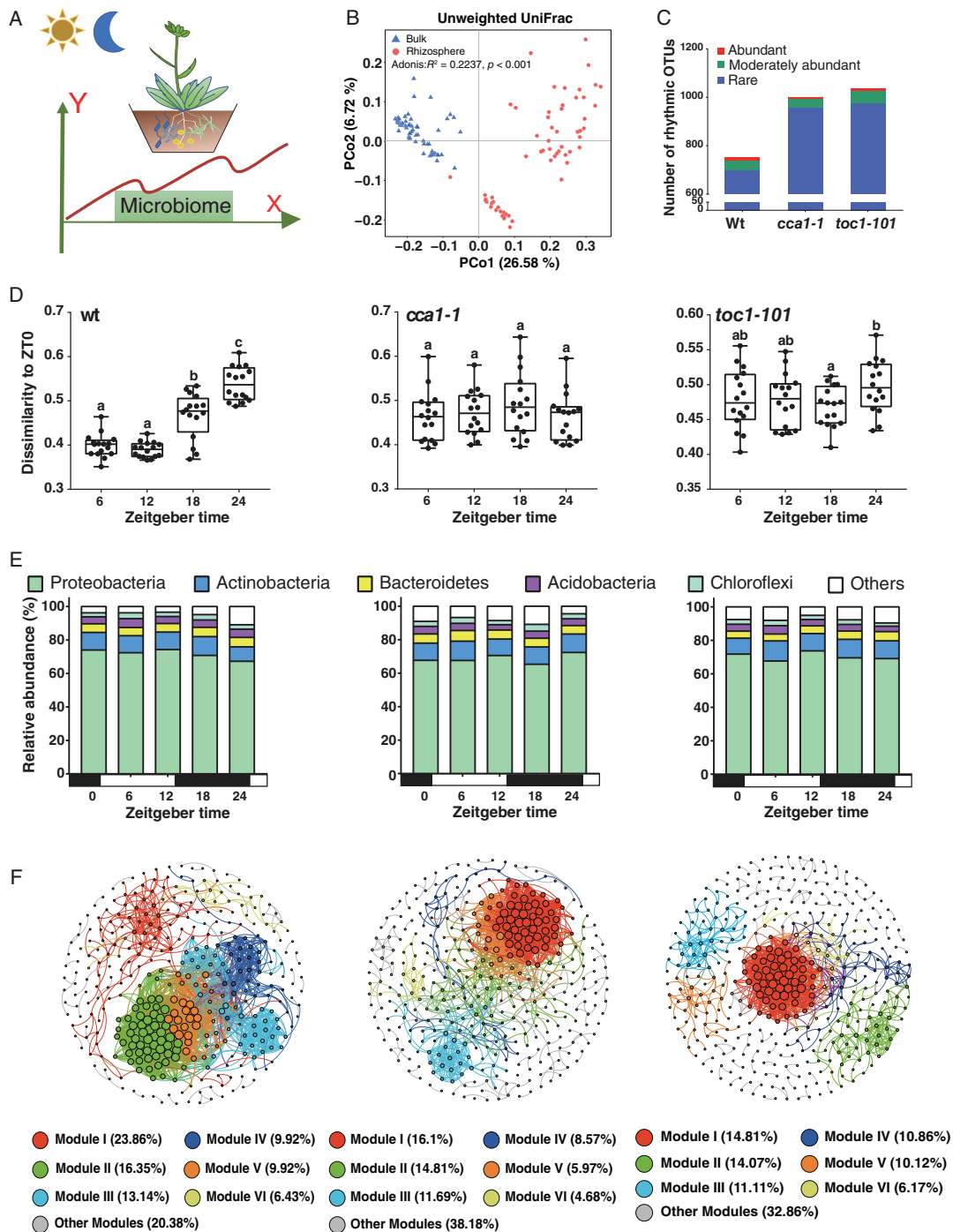


Fig. 3. *cca1-1* or *toc1-101* deletion alters diurnal change in bacterial abundances in the *Arabidopsis* rhizosphere.

A. Schematic of the diurnal patterns of the rhizospheric microbiome.

B. PCoA of the compositions of the microbial communities in the bulk soils and rhizospheres based on unweighted UniFrac distances.

C. Number of rhythmic OTUs in the wild-type (Wt) *Arabidopsis* and the two mutants; OTUs were divided into three classes (rare taxa, moderately abundant taxa and abundant taxa) described by Xue *et al.* (2018).

D. Dissimilarities based on unweighted UniFrac distances between the rhizospheric microbiomes at ZT0 and those at ZT6, ZT12, ZT18 and ZT24.

E. Fluctuation of the five most abundant phyla in the rhizospheres of wild-type *Arabidopsis* and the two mutants.

F. Co-occurrence networks of bacterial communities in the rhizospheres of wild-type *Arabidopsis* and the *cca1-1* and *toc1-101* mutants at the genus level. The lines represent strong (Spearman's $r > 0.8$ or < -0.8) and significant ($P < 0.05$) correlations calculated using the psych package in R. The size of each node is proportional to the number of connections (degree of each node calculated using Gephi). The nodes are coloured based on the modularity class determined using Gephi. Only the six most common modules for each genotype are shown.

(Supporting Information Fig. S10). Most of these rhythmic OTUs belonged to rare taxa, as described by Xue *et al.* (2018) (Fig. 3C). The dynamic change of the microbiomes during the 24-h period differed in the rhizosphere soil of the mutants and wild type based on unweighted UniFrac distances (Fig. 3D). The rhizosphere microbiome of the wild type was clearly coordinated with plant growth during the 24-h period as the dissimilarities to ZT0 continuously increased. The rhizosphere microbiomes for both the *cca1-1* and *toc1-101* mutants were more similar between ZT0 and the other timepoints, indicating that the rhizosphere microbiomes showed randomness and had no obvious temporal pattern during the 24-h period.

Proteobacteria and Actinobacteria were the two most abundant phyla of the rhizosphere microbiomes for all genotypes (Fig. 3E), consistent with the core root microbiome of *Arabidopsis* defined by Lundberg *et al.* (2012). The relative abundances of both Proteobacteria and Actinobacteria decreased for the wild type during the 24-h period. The relative abundance of Proteobacteria for the mutants fluctuated slightly during the 24-h period, but the relative abundance of Actinobacteria remained stable. The co-occurrence network analyses indicated that the rhizosphere microbes interaction was more complex and stronger in the wild type (average of 16.542 nodes) than the *cca1-1* (average of 9.321 nodes) and *toc1-101* (average of 7.96 nodes) mutants (Fig. 3F and Supporting Information Table S1).

Multiple linear regression analysis was performed to figure out the links between root exudates and rhizosphere microbes. Results showed that the root exudates contributed mainly to shaping the rhizosphere microorganisms (Figs. 4 and S11). There were 75 root exudates significantly regulated the relative abundance of specific microbes (Supporting Information Fig. S11 and Data set S2) and half of these exudates were lipids and lipid-like molecules.

Discussion

The composition of the rhizosphere microbial community of plants could be regulated by plant development (Chaparro *et al.*, 2014; Zhang *et al.*, 2018). This implies that plants select for specialized microbial communities that change in response to plant growth, and this selection is believed to be conducive to plant growth or reproduction (Chaparro *et al.*, 2014; Lu *et al.*, 2018; Zhong *et al.*, 2020). A previous study has reported an apparent discrepancy of the microbial community structure in the rhizosphere of *Arabidopsis* between the vegetative and bolting stage (Chaparro *et al.*, 2014), indicating that the microbial community of the rhizosphere is continually changing during this shift. Here, we explored whether

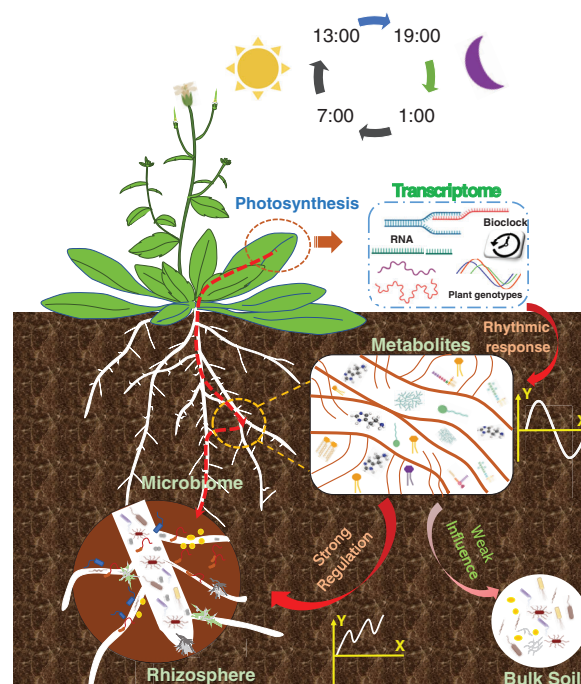


Fig. 4. Schematic of the role of the circadian clock of plants in determining the nature of the rhizospheric microbial community. A robust rhythmicity of genes and root exudates is an important driving force for the positive development of rhizospheric communities and thus plant development.

and how the plant circadian clock influences the rhizosphere microbial community during the transition from the vegetative to the reproductive phase.

The rhythmicity of gene transcription differed significantly between the wild type and mutant lines. The mutation of CCA1 and TOC1 did not reduce the number of cycling genes, but most cycling genes differed between the two mutants and the wild type. It seems like the mutation of the genes controlling the core circadian clock created another gene oscillating model in the mutants. The loss of genetic rhythmicity or the abnormality of its amplitude or phase in the *cca1-1* and *toc1-101* mutants was closely associated with the metabolism of lipids, carbohydrates, energy and amino acids. These metabolic pathways determine the synthesis of root exudates, which account for 10%–40% of the photosynthetically fixed carbon and ~15% of total plant nitrogen, including many kinds of primary and secondary metabolites (Haichar *et al.*, 2016), and act as key substrates or signalling molecules that affect microbial composition (Huang *et al.*, 2019; Ehlers *et al.*, 2020; Preece and Penuelas, 2020).

A previous study reported that the expression of many photosynthetic genes peaked near ZT4 (the light cycle in our study started 1 h after ZT0), when large amounts of sugar and other products of carbon fixation are

synthesized in *Arabidopsis* (Khan *et al.*, 2010). The exudates collected at ZT6 (continuously secreted by the roots from ZT0 to ZT6) would therefore benefit from the products of carbon fixation. In sharp contrast to ZT6, exudates at ZT0 were secreted at the end of the dark cycle when the energy source in the plant was nearly depleted. These metabolic discrepancies may account for the dissimilarity between the exudates collected at ZT0 and ZT6 in the wild type. The dynamic changes of root exudates in mutants differed from the wild type and were in accordance with the transcriptomic analysis. This indicates that plants can change the regularity of secretion for some exudates when mutations in the clock genes induced large-scale changes in gene transcription and subsequently affected metabolism. The rhythm of exudation in both mutants and the wild type, however, was notably less important than the rhythm of gene expression. This difference may have been due to (i) the joint control of metabolite synthesis by multiple genes, not all of which were rhythmically expressed; (ii) a time lag between gene expression and metabolite synthesis and secretion, or; (iii) the collection of exudates after 6 h of accumulation, which may have reduced the importance of the rhythm.

The root exudates contributed mainly to shaping the rhizosphere microorganisms. Half of these exudates were lipids and lipid-like molecules, which are critical exudates that are transferred from the plant to the microbes (Luginbuehl *et al.*, 2017; Siebers *et al.*, 2016) and play important roles in the communication between them. The lipid exudates differed between the mutants and the wild type mainly because of the abnormal rhythmicity of the genes involved in lipid metabolism. For example, the transcriptomic analysis indicated that the rhythmicity of genes involved in sphingolipid metabolism (*cca1-1*) and glycerolipid metabolism (*toc1-101*) varied more than the rhythmicity of these genes in the wild type. These two lipids are involved in establishing the membrane interface between plant and microbial cells (Siebers *et al.*, 2016). Besides, species interaction is a key driver of assemblage and dynamic changes of microbial community (Xiong *et al.*, 2020, Ratzke *et al.*, 2020). In addition to the rhythmic root exudates, the stronger species interactions, especially the positive one, in the wild type could also promote the highly dynamic development of the rhizosphere microbiome during the 24-h period (Badri *et al.*, 2010). This is reasonable, because if one species increases in abundance, the others will increase as well because of the positive feedbacks between them (Coyte *et al.*, 2015).

Staley *et al.* (2017) found that the composition of organic matter differed significantly between rhizosphere samples during a light and dark cycle and that 13% of the microbial community was cycling, suggesting that the rhizospheric bacterial community was regulated by the

circadian clock. We also found that some bacteria (14.43%–22.67%) in the rhizospheres of all genotypes had a diurnal rhythm. More interestingly, however, the entire microbial community in the rhizosphere of the wild type varied continuously from ZT0 to ZT24 rather than rhythmically. In contrast, the *cca1-1* and *toc1-101* mutants, which had abnormal circadian clocks, were weak drivers for adjusting the microbiome towards the 'designated' target. Similarly, the abundance of some rhizosphere microorganisms in barley is not rhythmic but fluctuates from light to dark timepoints (Nakamichi, 2011). The rhythmicity of genes may have led to the rhythmicity of some of the root exudates, but this rhythmicity could not be extrapolated to the microbial community. The rhythmic secretion of organic matter by plants would thus be beneficial for transforming the rhizosphere microbiome into a special community. This process is strictly programmed and controlled mainly by a biological rhythm. The disturbance of this rhythm or a change to a new oscillating model in the mutants would slow the dynamic fluctuation and development of the rhizosphere microbiome. We therefore hypothesize that the circadian clock in plants is an important driving force for the positive development of rhizosphere communities and thus plant development (Fig. 4), which has been neglected in most previous studies. In addition, although we think that the rhizosphere microbial community structure does not exhibit rhythm, this is still an open matter that whether the gene expression of the community shows rhythm. We would try to explore this interesting question using meta-transcriptome techniques in future studies.

Hubbard *et al.* (2018) found that the mutation of two clock genes (TOC1 and ZTL) had a strong impact on the *A. thaliana* rhizosphere microbiome, particularly the rare taxa. They also found that wild-type plants germinated earlier and were larger when inoculated with soils where the wild type had grown compared with genotypes with mutant clock genes. This finding implied that the microbial community in the rhizosphere of plants with a normal circadian clock would be beneficial to the germination and growth of their offspring. The role of the circadian clock is particularly important during flowering, and the mutation of core genes directly affects the timing of flowering (Nakamichi, 2011). Rhizosphere microorganisms can influence the timing of flowering (Lu *et al.*, 2018), so an abnormal clock may affect the rhizosphere microbes and ultimately indirectly affect flowering, in addition to the direct influence of the clock oscillator.

The compositions of the rhizosphere microbial communities of the two mutants and the wild type fluctuated with distinct patterns, indicating that the mutant genes of the core circadian clock in *Arabidopsis* led to different

programming effects on the rhizosphere microbial communities. The abundance of the core microbes for the wild type was regulated by rhythmic root exudation, especially the lipids and developed continuously due to the strong species interactions. This dynamic pattern may account for the effect of plant development on the composition of rhizosphere microbiome (Chaparro *et al.*, 2014).

Conclusions

Our study has made a step forward by connecting the circadian clock of the plant to the rhizosphere microbial composition and finding a strong influence. We obtained the detailed rhythmic characteristics of transcription and metabolism of wild-type and two mutant strains, and how they influenced the rhizosphere microbiomes. In this process, lipids were potential metabolites linking the plant host rhythmicity and rhizosphere microbiome assemblage. We do not know, though, the specific molecules dominating the regulation of microbial community by biological rhythms nor how does an abnormal microbial community generated by disturbed biological clock in turn affect plants. Future research will closely revolve around these two questions. Our results show that the robust rhythmicity of genes and root exudation due to circadian clock in plants is an important driving force for the positive succession of rhizosphere communities and thus plant development; however, such effect does not take the form of rhythm. Therefore, the composition of root microbes might be influenced by the plant circadian clock when environmental factors such as drought, global warming, light pollution or pesticide residues affect the plant endogenous diurnal rhythm. The variation of the composition of root microorganisms, as regulators of plant resistance and growth, very likely in turn influence the development and reproduction of the plants.

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Data availability

All data needed to evaluate the conclusions in the paper are present in the paper and/or the Supplementary Materials. The raw data for the transcriptomes and 16S rRNA

gene sequences were submitted to the NCBI Sequence Read Archive (SRA) database with the BioProject numbers PRJNA611909 and PRJNA611942, respectively.

Author contributions

H.Q., L.Z., Y-G.Z., and J.C. designed research; T.L., Z.Z. and Y.L. and H.C. performed research; Z.Z., Q.Z. and L.S. analysed data; and T.L., WJGM.P., J.P. and H.Q. wrote the paper.

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Supporting Information

Additional Supporting Information may be found in the online version of this article at the publisher's web-site:

Appendix S1. Supporting Information.

Data set S1. Root exudates identified from three *Arabidopsis* lines. The values were represented by the intensity of the peak.

Data set S2. Root exudates significantly correlated to the abundance of certain phyla or *Proteobacteria* classes. The values represented the standard regression coefficient calculated by the multiple linear regression in SPSS (v 20.0.0).