Understanding how nutrients affect gene expression will help us to understand the mechanisms controlling plant growth and development as a function of nutrient availability. Nitrate has been shown to serve as a signal for the control of gene expression in Arabidopsis. There is also evidence, on a gene-by-gene basis, that downstream products of nitrogen (N) assimilation such as glutamate (Glu) or glutamine (Gln) might serve as signals of organic N status that in turn regulate gene expression. To identify genome-wide responses to such organic N signals, Arabidopsis seedlings were transiently treated with ammonium nitrate in the presence or absence of MSX, an inhibitor of glutamine synthetase, resulting in a block of Glu/Gln synthesis. Genes that responded to organic N were identified as those whose response to ammonium nitrate treatment was blocked in the presence of MSX. We showed that some genes previously identified to be regulated by nitrate are under the control of an organic N-metabolite. Using an integrated network model of molecular interactions, we uncovered a subnetwork regulated by organic N that included CCA1 and target genes involved in N-assimilation. We validated some of the predicted interactions and showed that regulation of the master clock control gene CCA1 by Glu or a Glu-derived metabolite in turn regulates the expression of key N-assimilatory genes. Phase response curve analysis shows that distinct N-metabolites can advance or delay the CCA1 phase. Regulation of CCA1 by organic N signals may represent a novel input mechanism for N-nutrients to affect plant circadian clock function.

Nitrogen (N) is an essential nutrient and a metabolic signal that is sensed and transduced resulting in the control of gene expression in plants. Studies using nitrate reductase (NR) mutant plants have shown that nitrate can serve as a metabolic signal for inorganic N that regulates gene expression in Arabidopsis thaliana and other plant species (1–4). There is also ample though less direct evidence that the assimilated forms of N such as Glu or Gln may also serve as signals that regulate gene expression in plants (5, 6). The ability of plants to sense and respond to levels of inorganic and organic N-metabolites provides a mechanism to balance the availability of organic N resources within the plant with the need for N uptake. Because nitrate uptake, reduction, and its assimilation into organic form require energy, a mechanism that activates this N-assimilatory pathway based on sensing levels of organic N available in the plant is an efficient way to control N-use efficiency (3). In plants, the transcription of genes involved in the uptake and assimilation of inorganic N is induced when levels of organic N are low. Conversely, the uptake and reduction of inorganic N are shut off when levels of organic N are high (reviewed in ref. 7).

Recent microarray studies have shown that nitrate can cause changes in the expression of a large number of genes in Arabidopsis (1, 2). Treatment of Arabidopsis seedlings with low levels of nitrate has been shown to increase the levels of mRNA for hundreds of genes within minutes of exposure. The nitrate-responsive genes include nitrate transporters, NR and nitrite reductase, putative transcription factors, and stress responses genes, as well as genes whose products play roles in glycolysis, iron metabolism, and sulfate uptake (1, 2). In a related study, N-starved plants underwent a transcriptome/metabolome analysis 30 min and 3 h after nitrate treatment (4). The expression of nitrate transporters (at 30 min) preceded the induction of amino acid biosynthetic genes and the repression of amino acid breakdown genes (at 3 h). In addition, increases in amino acid levels were observed, consistent with the changes in expression of the cognate amino acid biosynthesis genes. Putative nitrate-responsive regulatory factors including transcription factors, protein kinases/phosphatases, and trehalose and hormone metabolic genes were also identified in that study. Recently, using a NR-null mutant, it was shown that nitrate, and not a product of nitrate reduction and assimilation, regulates the expression of genes involved in energy production, metabolism, glycolysis, and gluconeogenesis (1).

Nitrogen metabolism genes can be regulated by negative feedback of the products of N-assimilation. For example, the expression of the ammonium transporter gene ammonium transporter 1 (AMT1.1) is repressed in treatments with high levels of inorganic N. It has been shown that this repression is blocked by methionine sulfoximine (MSX), a nonmetabolizable analog of Glu that irreversibly inhibits glutamine synthetase and hence blocks N-assimilation into Gln (5). Thus, it appears that organic forms of N may regulate the uptake of N in plants. In addition, the genes encoding asparagine synthetase 1 (ASN1) and 2 (ASN2) are differentially regulated by organic and inorganic N sources. Organic N treatments were shown to positively regulate levels of ASN1 mRNA (6), whereas ASN2 gene expression appears to be responsive to inorganic N sources and not a downstream metabolite (1). Together, these studies prompt a model in which both inorganic and organic N sources can each regulate plant gene expression affecting N uptake, reduction, and assimilation.

In this study we used a genomic approach to identify gene networks whose expression is regulated by Glu or Glu-derived metabolites (organic N) in plants. Network analysis of the genes that responded to organic N revealed that transcription control of gene expression is important for a subnetwork of metabolic genes involved in the synthesis and degradation of asparagine (Asn). The metabolic network discovered in this analysis provides molecular evidence that the assimilation of organic N by plants is a major route to regulate gene expression affecting N uptake, reduction, and assimilation.


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lecular evidence for regulation of N-use at the level of gene expression. Moreover, the transcription factors regulated by organic N associated with this network provide a mechanistic link between circadian clock function and N-assimilation in plants.

Results

Inorganic Versus Organic N Responses. To uncouple gene responses to inorganic N from those elicited by downstream products of inorganic N-assimilation we performed treatments of Arabidopsis seedlings with combinations of inorganic N (nitrate and ammonium), organic forms of N (e.g., Glu and Gln), and MSX, an inhibitor of glutamine synthetase (8) [supporting information (SI) Fig. 5]. Genes regulated by inorganic N signals should be unaffected by MSX treatment. By contrast, genes responding to a downstream organic N signal should fail to show induction by inorganic N treatments if Glu/Gln synthesis is blocked by MSX. This block of induction by MSX should be relieved by Glu treatment. Following this rationale, 2-week-old Arabidopsis seedlings grown on low concentrations of N (1 mM NO3−) were transferred to media containing 40 mM NO3− and 20 mM NH4+ (referred to as “Nms”). Seedlings were then harvested after a 2-h treatment time. This treatment was carried out alone (Nms) or in the presence of 1 mM MSX (Nms+MSX) or 1 mM MSX and 10 mM Glu (Nms+MSX+Glu). The Nms treatment consists of the same N source found in standard MS salts, which is the established standard amount of N for plant growth (9). A concentration of 1 mM MSX has previously been established as effective in blocking the N repression of AMT1.1 in Arabidopsis seedlings and in decreasing levels of internal organic N (5). We chose a concentration of 10 mM for Glu treatments because this has been shown to be effective in the regulation of N-assimilatory genes while not being high enough to be detrimental to plant growth or development (6). To evaluate the effect of MSX alone, plants were exposed to growth media that contained MSX. To control for the effect of the plant transfer to distinct media, plants were transferred onto media plates without any of the treatment factors. This latter control was used as the baseline for the microarray experiments described below.

To evaluate our experimental design, we determined the mRNA level of genes shown to be responsive to organic N (ASN1) or inorganic N (ASN2) by reverse transcription followed by real-time quantitative PCR (RT-qPCR). This analysis showed that the ASN1 mRNA level was induced 3.5-fold by the Nms treatment as compared with the control (SI Fig. 6A, compare Nms to the control). This induction of ASN1 mRNA was blocked when MSX was present (SI Fig. 6A, Nms+MSX versus Nms). Importantly, when exogenous Glu or Gln was added, ASN1 levels were induced regardless of the presence of MSX (SI Fig. 6A, compare Control to Nms+MSX+Glu and Nms+MSX+Gln). These results indicate that the induction of ASN1 is due to Glu or a downstream metabolite, as shown previously (10, 11). In addition, the control treatments showed that MSX alone does not induce expression of ASN1 or ASN2 (SI Fig. 6). The addition of Glu or Gln partially blocked the induction of ASN2 by the Nms treatment (SI Fig. 6B), consistent with the negative regulation by amino acids seen previously (10). Conversely, the induction of ASN2 mRNA by Nms was insensitive to MSX addition (SI Fig. 6B), suggesting that the induction was mediated by an inorganic N source. This finding was consistent with previous data, which indicate that ASN2 gene expression correlates with ammonium levels (12).

Global Genomic Responses to Organic and Inorganic N Signals. To investigate global gene expression changes that are mediated by Glu or a Glu-derived metabolite, we monitored the plant transcriptome using the ATH1 Affymetrix gene chip. Total RNA was extracted from plants treated with Nms, Nms+MSX, or Nms+MSX+Glu as described above; two biological replicates per treatment were performed. The Nms+MSX+Gln treatments were not analyzed with microarrays because Gln and Glu responses were similar in our hands (SI Fig. 6 and data not shown). RNA was labeled and hybridized to the microarrays, the raw intensity values were normalized, and the data were filtered as described in Materials and Methods. A gene was kept in the data set only if its expression was reproducible and reliable across the three different treatments (Nms, Nms+MSX, and Nms+MSX+Glu). A total of 5,904 genes were identified that passed these stringent quality control criteria. To verify the microarray results, we analyzed the mRNA levels of selected genes by RT-qPCR including the TAZ zinc binding (At4g37610) and bZIP (At5g49450) transcription factors with results similar to the microarray data (SI Fig. 7). As expected, the genomic experiments verified the previous observation that AMT1.1 and ASN1 are regulated by organic N. In addition, our results identified additional genes regulated by organic N as described below.

Genes were categorized based on their response to the treatments using the Affymetrix change calls: induced (I), no change (NC), or decreased (D). Each gene was assigned a three-part code (e.g., I-NC-I), which corresponds to the gene expression response in the Nms, Nms+MSX, and Nms+MSX+Glu treatments, respectively. We found 21 unique patterns of response (SI Table 2). The largest pattern was NC-NC-NC, representing 5,070 genes not affected by the treatments. We grouped the remainder 20 patterns (834 genes) into six classes that summarize the types of N responses observed: class A, inorganic N; class B, inorganic N with Glu feedback; class C, organic N with no exogenous Glu rescue; class D, exogenous Glu; class E, exogenous and endogenous Glu; and class F, exogenous and endogenous Glu with opposite effects (Table 1).

We compared the genes regulated in our experiments to published results (1). This previous study identified 595 genes that responded similarly to nitrate treatment in both a NR-null mutant and wild-type plants. Because the mutant plants cannot assimilate nitrate, the responses observed were attributed to the action of nitrate as a signal and not a downstream metabolite. Eighty of these 595 genes showed consistent and reliable responses in our experiments. Surprisingly, only 17 of these genes were found regulated by inorganic N signals in both studies (SI Table 3). Among these, we

<table>
<thead>
<tr>
<th>Table 1. N responses</th>
<th>Treatment</th>
<th>Nms</th>
<th>MSX</th>
<th>Nms+MSX</th>
<th>Glu</th>
<th>Genes per class</th>
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<tr>
<td>Response</td>
<td></td>
<td></td>
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<tr>
<td>Class A, inorganic N</td>
<td>D</td>
<td>D</td>
<td>D</td>
<td>100</td>
<td>159</td>
<td></td>
</tr>
<tr>
<td>Class B, inorganic N and Glu</td>
<td>D</td>
<td>D</td>
<td>NC</td>
<td>30</td>
<td>48</td>
<td></td>
</tr>
<tr>
<td>Class C, internal Glu</td>
<td>D</td>
<td>NC</td>
<td>NC</td>
<td>194</td>
<td>334</td>
<td></td>
</tr>
<tr>
<td>Class D, external Glu</td>
<td>NC</td>
<td>NC</td>
<td>D</td>
<td>89</td>
<td>126</td>
<td></td>
</tr>
<tr>
<td>Class E, internal/external Glu</td>
<td>NC</td>
<td>D</td>
<td>NC</td>
<td>36</td>
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<tr>
<td>Class F, opposite internal/external Glu</td>
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<td>NC</td>
<td>I</td>
<td>2</td>
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<td>Total</td>
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Patterns of expression based on their response to the Nms, Nms+MSX, and Nms+MSX+Glu treatments. Genes were categorized into six classes based on these patterns. D, decreased; I, induced; NC, no change.
found nitrite reductase and several high-affinity nitrate transporters. The majority, 58 of these 80 genes (73%), belonged to the C, D, or E classes in our studies, suggesting that many previously described nitrate-responsive genes may respond to organic N signals (SI Table 3).

**N Signals Control Amino Acid Metabolism in Arabidopsis Seedlings.** To evaluate the biological significance of the observed patterns of response to the treatments, we analyzed the distribution of functional categories in the six classes defined in Table 1 using the BioMaps program (13). To focus on the most prominent biological processes affected, we analyzed overrepresented functional terms (P ≤ 0.01) with five or more genes (SI Table 4). This analysis indicated that inorganic N represses amino acid biosynthesis and, in particular, a subset of genes related to the metabolism of S-containing amino acids. In addition, increased levels of internal Glu induced the expression of genes involved in cell wall biosynthesis, especially genes in the xyloglucan:xyloglucosyl transferase family. Internal Glu also repressed genes involved in several aspects of metabolism, most prominently amino acid and carbohydrate metabolism. Internal and external organic N also repressed genes involved in secondary metabolism. These results indicate that the balance between organic and inorganic N controls the expression of genes involved in N-reduction, N-assimilation, and amino acid metabolism in Arabidopsis plants and coordinates N-assimilation with cellular processes including, for example, cell wall biosynthesis.

**Network Analysis Reveals a Metabolic Gene Network Connected to Regulatory Transcription Factors Regulated by Organic N.** To uncover the mechanism underlying gene regulation in response to sensing Glu or a Glu-derived product, we used network analysis to identify the subnetwork of genes regulated by organic N (Fig. 1). We generated the subnetwork of N-regulated genes using an Arabidopsis multinetwork described previously (13). Cytoscape was used to visualize the resulting subnetwork wherein genes were represented as nodes connected by edges that represented distinct interactions (e.g., metabolic reactions and regulatory interactions). In addition to the interactions described previously (13), we predicted regulatory connections between genes and associated transcription factors (Materials and Methods). To identify putative “master regulators” that control the expression of genes regulated by organic N, the transcription factors regulated in these experiments were ranked based on the number of regulatory connections in the subnetwork (SI Table 5). At the top of the list were a Myb family transcription factor (At1g74840), the central clock gene CCA1 (At2g46830), and a golden 2-related transcription factor (GLK1; At2g20570). Interestingly, both CCA1 and GLK1 were predicted to positively affect the expression of a gene for glutamine synthetase (GLN1.3) (which uses Glu in a biosynthetic reaction) and to negatively affect the expression of a glutamate dehydrogenase gene (GDH1) (which catalyzes Glu (Fig. 1). Moreover, our analysis suggests that both CCA1 and GLK1 block the expression of a bZIP transcription factor (bZIP1), which is predicted to induce the expression of the Glu-dependent ASN1 gene. Thus, Glu regulation of the GLK1 and CCA1 transcription factors appears to coordinate the expression of genes involved in making Glu (GLN1.3) versus those involved in metabolizing Gln into Glu (ASN1 and GDH1). In addition, another gene in this gene subnetwork encodes a putative asparaginase gene (ASN) that controls the degradation of Asn (to Asp and Glu) (Fig. 1). The ASN1 (Asn biosynthesis) and ASN (Asn degradation) genes had inverse expression patterns (correlation = −0.51), suggesting that Glu coordinates the reciprocal regulation of Asn synthesis and degradation by coordinating an antiregulation of the cognate.

**Validation of Network Model Predictions Highlights the Regulatory Role of CCA1 in the N-Assimilatory Pathway.** The model in Fig. 1 predicts that CCA1 and/or GLK1 genes are important regulators of genes involved in N-assimilation and that overexpression of either one of these genes would repress the expression of ASN1 and GDH1 and induce the expression of the GLN1.3 gene. Conversely, a knockout of the CCA1 or GLK1 gene should increase ASN1 and GDH1 expression levels and diminish GLN1.3 mRNA levels. To test these hypotheses, we used previously characterized CCA1 overexpressor (CCA1-over) (14) and GLK1 gene knockout (glk1) (15) lines. We anticipated a stronger phenotype for the overexpressor lines as compared with the knockout, because the model predicts...
The mean normalized to clathrin (At4g24550). The mean compared with wild type in the predictions of the model, were decreased in the GDH1 ASN1 CCA1 protein to a region of the we demonstrated that the ChIP assays could detect binding of we used ChIP assays using a CCA1 antibody (Fig. 2 expression of the target genes will be direct. To test this hypothesis, the mRNA level was also repressed in the whole seedlings, and RT-qPCR was performed to determine mRNA levels for GDH1, ASN1, GLN1.3, and GDH1 in the three genotypes. As shown in Fig. 2, all three genes tested showed altered expression patterns in the mutant lines used (as determined by analysis of variance, P ≤ 0.05), which were consistent with the predicted network model shown in Fig. 1. In addition, bZIP1 mRNA level was also repressed in CCA1-ox (SI Fig. 8). ASN1, GDH1, and GLN1.3 mRNA levels were not altered in the glk1 line, with the exception of a small increase in GDH1 mRNA levels. This is probably because of the redundant function of GLK1 and CCA1 in regulating the expression of the tested genes. In contrast, and as predicted by the model shown in Fig. 1, ASN1 and GDH1 levels were decreased in the CCA1-ox line. Also consistent with the predictions of the model, GLN1.3 mRNA levels were increased as compared with wild type in the CCA1-ox.

The network model predicts that the effect of CCA1 on the expression of the target genes will be direct. To test this hypothesis, we used ChIP assays using a CCA1 antibody (Fig. 2B). As controls, we demonstrated that the ChIP assays could detect binding of CCA1 protein to a region of the TOCl promoter, a known target of CCA1, but was not able to detect the ZTL promoter, which has no circadian oscillation at the mRNA level. Consistent with the model for CCA1, ChIP assays in both wild-type and CCA1-ox lines were able to confirm binding of CCA1 to the promoter regions of GLN1.3, GDH1, and bZIP1 promoters. These results support the model and indicate that CCA1 regulates expression of bZIP1, GDH1, and GLN1.3 genes directly.

N-Nutrient Signals Act as Input to the Arabidopsis Circadian Clock. CCA1 is a key component of a negative feedback loop at the center of the Arabidopsis circadian clock (16, 17). Because our results showed that N treatments affected CCA1 expression, we hypothesized that N might serve as an input capable of affecting the circadian clock function. To test this hypothesis, we provided pulses of inorganic or organic N at intervals spanning a circadian cycle and determined the effects on the phase of the oscillation in CCA1::LUC expression. Each treatment resulted in stable phase shifts indicating that N status serves as an input to the circadian clock (Fig. 3 and SI Fig. 9). Inorganic N and 10 mM Glu treatments conferred slight phase advances whereas 10 mM Gln conferred only delays. The Nms and Glu pulses did not affect the period, but the Gln pulse shortened the period as determined by one-way analysis of variance and Dunn’s multiple-comparison tests. Thus, the clock regulates a number of steps in N metabolism, such as NR expression and activity (18) and ASN1 expression (this and previous studies (19)). In turn, N status feeds back to the clock, at least in part through its effect on CCA1 expression.

Discussion

We used genomic and pharmacological approaches to distinguish organic from inorganic N responses in Arabidopsis seedlings. The majority of the genes regulated by the N treatments used in our study (81%) responded to organic N signals. Among the genes regulated by organic N, we distinguished two classes of genes: (i) genes that responded only to external Glu application and (ii) genes that responded to internal and external sources of Glu (Table 1). The difference in these two expression patterns raises the possibility that there are different mechanisms for sensing internally produced cellular Glu versus extra cellular Glu. The other possibility is that the differences in internal versus external Glu responses observed in our study reflect distinct threshold responses to Glu levels. There is precedence for internal and external Glu-sensing mechanisms in other organisms. Bacteria regulate ammonium assimilation via a mechanism involving PII, a sensor that measures levels of α-keto-glutarate and Glu (20). Plants contain a PII protein that is localized to chloroplasts (21), a potential sensor of internal levels of Glu. By contrast, extracellular Glu is sensed by Glu receptors in animal
The phase shifts in response to tested with our 4-h N pulses. Alternatively, N treatment may also activity changes from reduced mRNA, at least over the time frames suggests that posttranscriptional regulation buffers against CCA1 is not definitively established, they may involve induction of in the night (29). Although the molecular basis of these phase shifts back to the clock, at least in part through its effect on received input not only from environmental stimuli but also from emerging view of the circadian clock is as a key integrator of multiple metabolic pathways, many of which are subject to circadian regulation. Thus, the clock regulates a number of steps in N assimilation in plants. Our data are consistent with Asn serving as a major N storage compound is optimized and regulated at the level of transcription by increasing levels of ASN1 and decreasing levels of ANS transcripts. Our results are consistent with Asn signaling for regulation of N-assimilation in Arabidopsis. We validated our network model by measuring mRNA levels of the target genes in CCA1-ox and glk1 knockout lines. As predicted, ASN1 and GDH1 mRNA levels were down-regulated and GLN1.3 mRNA was elevated in the CCA1-ox line. In contrast, mRNA levels for these three genes were not affected in the glk1 knockout line. The lack of a molecular phenotype in the glk1 knockout may be explained by the fact that CCA1 and GLK1 are predicted to have the same regulatory function in the subnetwork. We also predicted the regulations that CCA1 directly targets a number of genes in the network using CCA1 antibodies in ChIP experiments. Because CCA1 is one of the central components of the circadian clock in Arabidopsis, regulation of CCA1 expression in response to organic N suggests that the circadian clock may receive N-nutritional inputs in plants. Thus, in addition to light and temperature (17, 25), nutrients—such as N—are used as input for the clock. Our phase response curve analysis results are consistent with weak (type 1) resetting similar to those observed in response to light pulses in Lemma gibba (26), KCl, or ethanol pulses in Phaseolus coccineus (27) and CAMP or imidazol pulses in Trifolium repens (28), strengthening the hypothesis that N status feeds back to the clock, at least in part through its effect on CCA1 expression. In Arabidopsis, light pulses evoke strong delays (~8 h) in the early night and strong advances (5–10 h) later in the night (29). Although the molecular basis of these phase shifts is not definitively established, they may involve induction of CCA1 by light (14). In our experiments, N treatment would decrease CCA1 mRNA abundance. That this elicits only small phase shifts suggests that posttranscriptional regulation buffers against CCA1 activity changes from reduced mRNA, at least over the time frames tested with our 4-h N pulses. Alternatively, N treatment may also modulate other clock components at the mRNA, protein abundance, or protein activity level in ways that reduce the magnitude of the phase shifts in response to CCA1 mRNA decrease. The emerging view of the circadian clock is as a key integrator of multiple metabolic and physiologic processes (19, 30). As such it receives input not only from environmental stimuli but also from multiple metabolic pathways, many of which are subject to circadian regulation. Thus, the clock regulates a number of steps in N metabolism, such as NR expression and activity (18) and ASN1 expression [this and previous studies (19)]. In turn, N status feeds back to the clock, at least in part through its effect on CCA1 expression. This feedback is more subtle than the effects of saturating light pulses, and our results are consistent with N status fine-tuning clock function rather than conferring large changes such as those observed in response to light (Fig. 4). Oscillations in the mRNA of genes that code for metabolic enzymes could have an impact on metabolic levels, as recently shown (31). Predicting time of food availability is key for the survival in most animals (32). Our data suggest that this may also be the case in Arabidopsis, e.g., anticipating the availability of carbon skeletons produced by photosynthesis to assimilate inorganic N into amino acids. Moreover, our data provide a plausible molecular mechanism for how this could happen in plants via CCA1. Our study thus provides evidence that plant nutrition, like in animals, is tightly linked to circadian functions as previously hypothesized (19). Recently, it was shown that the central clock gene Per2 is necessary for food anticipation in mice (33). Our data indicate that the central clock gene CCA1 plays a role in circadian regulation of N-assimilation in plants. Our data are consistent with a model in which the N-assimilatory pathway is a downstream target of the clock with CCA1 being the direct regulatory factor. Moreover, Glu or other Glu-derived signals act as input to the circadian clock providing a link between plant N-nutrition and circadian rhythms (Fig. 4).

Materials and Methods

Plant Growth Conditions. Wild type A. thaliana Columbia-0 strain was used in all experiments unless indicated otherwise. Seeds were sterilized as described (34) and sown onto basal MS salts (custom-made; Gibco) with 0.5% (wt/vol) sucrose, 0.8% BactoAgar, and 1 mM KNO3. After 14 days under long-day (16/8-h light/dark) conditions at 22°C, plants were transferred for 2 h in the light at the start of their light cycle by transferring them to basal medium with 0.5% sucrose and a combination of inorganic N sources (20 mM KNO3 and 20 mM NH4NO3 with or without 1 mM MSX (M-5375; Sigma), 10 mM Glu (G-1501; Sigma), and/or 10 mM Gln (G-3126; Sigma)).

RNA Isolation and RT-qPCR. RNA was isolated from whole plants with the TRIzol reagent and according to the instructions of the manufacturer (15596; Invitrogen). cDNA synthesis from whole mRNA extractions was carried out according to the kit manufacturer’s instructions (11146-024; Invitrogen). RT-qPCR was carried out with a LightCycler (Roche Diagnostics) as described (35).

Microarray Experiments and Analysis. cDNA synthesis, array hybridization, and normalization of the signal intensities were performed according to the instructions provided by Affymetrix. Raw data were processed with MASt5.0 software. Each hybridization was normalized to a median intensity of 150. Each treatment replica was compared with the two baselines to generate four comparisons per treatment. Data points with absent/marginal calls (Affymetrix quality control) in both baseline and treatment were removed. Data points with absent call in one hybridization and present call in the other hybridization were eliminated if the probe called present had a signal intensity of <100. We then summarized the response of each gene using the Affymetrix change calls “1” for increased, “-1” for decreased, and “NC” for no changed. Data points were considered only if the...
change calls were consistent in at least three of the four comparisons. This filtering resulted in a set of 834 genes that were detected and responded consistently in our experiments. We used custom-made S-PPLUS and PERL functions to analyze the data.

**Network Analysis.** For network analysis we used our existing network model of gene plant interactions (13). In addition, we predicted protein:DNA interactions as follows: The consensus sequence for transcription binding sites from curated databases (36, 37) were searched in 1,500 bp of upstream sequence using the DNA pattern search tool from the RSA tools server with default parameters (38). The search was performed in both strands of DNA, the upstream region was not allowed to overlap with the coding region of the upstream gene, and motif matches were not allowed to overlap. A motif was considered overrepresented if it was present in an upstream sequence more than three times the standard deviation above the mean occurrence in all of the upstream sequences in the genome. A protein:DNA interaction was predicted when the upstream sequence of a gene contained an overrepresentation of the regulatory motif for that transcription factor and the expression of the transcription factor and putative target gene was highly ($\geq 0.7$) or less than or equal to $-0.7$) and significantly ($P \leq 0.01$) correlated. These predictions are available in SI Data Set 1. Similar regulatory motif predictions for other microarray data sets can be generated with the VirtualPlant system (www.virtualplant.org) using the “Gene Networks” tool.

**Chip Assays.** Immunoprecipitations (IP) were performed as previously described (39). Briefly, 2-week-old wild-type and CCA1-oX plants were collected at the beginning of the light cycle and immediately fixed in 1% formaldehyde for 15 min in a vacuum at room temperature. Cross-linking was stopped by the addition of glycine to a final concentration of 0.125 M. Nuclei were prepared for chromatin isolation. The isolated chromatin was sonicated 10 times for 20 s each at 100% power (Diagenode Bioruptor) in an ice-water bath. A small aliquot of sheared chromatin was removed to serve as control. The diluted chromatin was used for IP with the CCA1 antibody and one control IP without antibody. SI Table 6 lists primer sequences used for amplification of the CCA1 binding sites in each of the genes tested. PCR amplifications included 95°C for 2 min followed by 36 cycles of 95°C for 15 s, 58°C (for bZIP1, TOC1, and GDH1) or 60°C (for ZTL and GLN1.3) for 30 s, and 72°C for 30 s.

**Circadian Phase Response Curves.** CCA1::LUC seedlings were entrained on MS basal medium plus 0.5% sucrose and 1 mM KN03 for 8 days in 16:8 hr light/dark (100 - 150 μmol m$^{-2}$ s$^{-1}$), after which seedlings were moved into continuous light. At 3-h intervals, seedlings ($n = 16$ per treatment) were transferred to fresh solid medium plus 2 ml of liquid NMS or medium containing 10 mM Glu or 10 mM Gin for 4 h, then rinsed in liquid entrainment medium three times for a total of 30 min and transferred individually to the wells of 36-well microtiter plates containing fresh solid media for luciferase activity measurements, which were determined with a Packard TopCount scintillation counter as described (40). The period and phase of rhythms after the pulses were determined by fast-Fourier transform nonlinear least-square analysis (41). The phase shifts were calculated as described (30).

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