



Evolution of Daily Gene Co-expression Patterns from Algae to Plants

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Specialty section:

This article was submitted to
Plant Evolution and Development,
a section of the journal
Frontiers in Plant Science

Received: 25 January 2017

Accepted: 28 June 2017

Published: 13 July 2017

Citation:

de los Reyes P, Romero-Campero FJ,
Ruiz MT, Romero JM and Valverde F
(2017) Evolution of Daily Gene
Co-expression Patterns from Algae to
Plants. *Front. Plant Sci.* 8:1217.
doi: 10.3389/fpls.2017.01217

Daily rhythms play a key role in transcriptome regulation in plants and microalgae orchestrating responses that, among other processes, anticipate light transitions that are essential for their metabolism and development. The recent accumulation of genome-wide transcriptomic data generated under alternating light:dark periods from plants and microalgae has made possible integrative and comparative analysis that could contribute to shed light on the evolution of daily rhythms in the green lineage. In this work, RNA-seq and microarray data generated over 24 h periods in different light regimes from the eudicot *Arabidopsis thaliana* and the microalgae *Chlamydomonas reinhardtii* and *Ostreococcus tauri* have been integrated and analyzed using gene co-expression networks. This analysis revealed a reduction in the size of the daily rhythmic transcriptome from around 90% in *Ostreococcus*, being heavily influenced by light transitions, to around 40% in *Arabidopsis*, where a certain independence from light transitions can be observed. A novel Multiple Bidirectional Best Hit (MBBH) algorithm was applied to associate single genes with a family of potential orthologues from evolutionary distant species. Gene duplication, amplification and divergence of rhythmic expression profiles seems to have played a central role in the evolution of gene families in the green lineage such as *Pseudo Response Regulators (PRRs)*, *CONSTANS-Likes (COLs)*, and *DNA-binding with One Finger (DOFs)*. Gene clustering and functional enrichment have been used to identify groups of genes with similar rhythmic gene expression patterns. The comparison of gene clusters between species based on potential orthologous relationships has unveiled a low to moderate level of conservation of daily rhythmic expression patterns. However, a strikingly high conservation was found for the gene clusters exhibiting their highest and/or lowest expression value during the light transitions.

Keywords: daily rhythmic genes, evolution, co-expression networks, systems biology, circadian, *Arabidopsis*, *Chlamydomonas*, *Ostreococcus*

INTRODUCTION

The evolution of the green lineage of photosynthetic organisms from unicellular green algae to land plants is subject to intense research. Recently, genomic analysis have been applied to study key events in the evolution of the green lineage such as terrestrialization (Delwiche and Cooper, 2015; de Vries et al., 2016). Nevertheless, the integration of genomics and transcriptomics to study the

evolution of gene expression patterns in the green lineage has only recently been explored (Romero-Campero et al., 2013; Ruprecht et al., 2017). Photosynthetic organisms are particularly influenced by daily light/dark transitions as their main energy income is a very demanding light-dependent process. Therefore, those plant ancestors (whose extant representative species are unicellular green algae) that were able to anticipate daily fluctuations and schedule their physiological processes accordingly, evolved into the present plant species (Mora-García et al., 2017). Transcriptomics have become a powerful tool to study the global influence of daily light/dark cycles in photosynthetic organisms. Thus, in higher plants, it has been described that between a third and a half of their gene expression is regulated by the circadian clock (Covington et al., 2008; Michael et al., 2008), whereas in algae between 80 and 90% of the transcriptome follows light-dependent daily rhythmic patterns (Monnier et al., 2010; Zones et al., 2015). It seems then that daily rhythmic regulation of gene expression reached a maximum in algae and has substantially decreased in Spermatophytes. Key biological processes such as starch and sugar metabolism exhibit daily rhythmic patterns in both plants and algae (Bläsing et al., 2005; Sorokina et al., 2011), but these analysis remain fragmented and focused on individual species. In order to better understand the evolution of daily rhythmic gene expression between algae and plants, confirm the tendency to decrease daily regulation and determine the evolution of daily regulated biological processes from algae to plants, a series of comparative Systems Biology analysis integrating genomics and transcriptomics has been used in this study.

The availability of massive amounts of transcriptomic data obtained from different species under equivalent environmental conditions has enabled the use of comparative transcriptomics methodologies to study the evolution of key biological processes (Trachana et al., 2010). In this work, database available RNA-seq and microarray data generated over 24 h periods in neutral days (ND: 12 h of light/12 h of dark) and long days (LD: 16 h of light/8 h of dark) conditions from three different model photosynthetic species (*Arabidopsis thaliana*, *Chlamydomonas reinhardtii*, and *Ostreococcus tauri*) have been integrated and analyzed. These photosynthetic eukaryotes represent distant phylogenetic groups. Among the *Chlorophyta* algae division, the marine Prasinophyceae microalga *Ostreococcus* is considered a representative of ancient microalgae, one of the smallest eukaryote (picoeukaryote) and constitutes an important part of sea phytoplankton (Derelle et al., 2006; Palenik et al., 2007). Due to its small genome (13 Mb), whose sequencing has recently been improved (Blanc-Mathieu et al., 2014), its non-flagellar small body (0.8 μm), its single copy organella (Henderson et al., 2007) and its planktonic life style, it has been considered as the smallest possible photosynthetic eukaryote (Raven et al., 2013) and promising organism for Systems Biology (Thommen et al., 2015). *Chlamydomonas* is a Chlorophyceae microalga and has been used as a model for photosynthetic organisms' studies for many years (Harris, 2001). *Chlamydomonas* has two polar flagella (10 μm body), a much bigger genome than *Ostreococcus* (120 Mb; Merchant et al., 2007) and lives in sweet water environments. Omics are currently intensely used to explore *Chlamydomonas*

potential in biotechnological applications (Aucoin et al., 2016). As a representative Spermatophyte or seed plant, *Arabidopsis* has several characteristics that make it especially useful for Systems Biology studies in general and circadian experiments in particular, as well as a bigger genome and a complex physiology in comparison to microalgae (Van Norman and Benfey, 2009; Koornneef and Meinke, 2010).

Arabidopsis has been extensively used as a model to describe the basic aspects of plant circadian clock regulation over daily rhythms (Millar, 2016; Nohales and Kay, 2016). In a few words, it is formed by three interlocked positive/negative feedback loops. CIRCADIAN CLOCK-ASSOCIATED 1 (CCA1) and LATE ELONGATED HYPOCOTYL (LHY) myb transcription factors constitute the positive/negative "morning loop" together with PRR9, PRR7 and PRR5 (Nakamichi et al., 2012; Liu et al., 2013, 2016; Kamioka et al., 2016). The negative "central loop" is constituted by CCA1/LHY repression over the gene *TIME OF CAB EXPRESSION 1 (TOC1)* that, in time, represses the "morning loop" genes and activates evening-regulated genes in a 24 h-long feed-back loop (Huang et al., 2012). Over this basic core, the circadian genes: *GIGANTEA (GI)*, *ZEITLUPE (ZTL)*, *EARLY LIGHT FLOWERING 3 (ELF3)*, *ELF4*, *LUX ARRHYTMO (LUX)* among others, constitute the "evening loop" that refines the clock and allows for the complex response to the changing daily conditions, including light and temperature inputs (Miyazaki et al., 2015). Outputs of this clock are cell wall synthesis, photosynthetic and starch metabolism genes, among many others (Adams and Carré, 2011). Taking *Arabidopsis* as model, a much simpler clock has been described in *Chlamydomonas* and *Ostreococcus*, where only some genes of the higher plant circadian clock have been identified so far (Mittag et al., 2005; Corellou et al., 2009). In this way, most evolutionary studies have focused on phylogenetic analyses of the key genes regulating the circadian clock (Serrano-Bueno et al., 2017) whereas the analysis of global rhythmic patterns conservation and evolution among different photosynthetic species still remains to be explored.

In this work, phylogenomic and transcriptomic data integration and analysis have been performed by gene co-expression networks construction (Romero-Campero et al., 2013; Gehan et al., 2015; Ruprecht et al., 2017) and a novel algorithm for the identification of potential orthologues called Multiple Best Bi-directional Hit (MBBH). Using clustering techniques, specific gene clusters or modules that showed a rhythmic daily regulation have been identified. Interestingly, these clusters consist of groups of highly co-expressed genes involved in particular biological processes such as cell cycle progression, photosynthesis and ribogenesis, revealing a significant temporal organization in their specific gene co-expression patterns. By comparing the gene modules identified in the different gene co-expression networks obtained for each species, it was possible to determine which biological processes have conserved a daily rhythmic co-expression pattern over the green lineage and which ones have evolved into different patterns. Additionally, a web based software tool, CircadiaNET (<http://viridiplantae.ibvf.csic.es/circadiaNet/>) has been developed that will allow researchers to independently

analyze their circadian genes of interest studying the biological processes they are potentially involved in, the conservation or evolution of the gene co-expression patterns they follow, as well as the transcription factor (TF) binding sites that are significantly present in their promoters.

MATERIALS AND METHODS

Identification of Putative Orthologous Proteins

The protein sequences of the three photosynthetic species analyzed in this study were downloaded from publicly available databases. *Chlamydomonas* v5.5 and *Arabidopsis* TAIR10 proteins were downloaded from Phytozome (<http://www.phytozome.net/>) (Goodstein et al., 2012), while *Ostreococcus* proteins v2 were downloaded from ORCAE (<http://bioinformatics.psb.ugent.be/orcae/>) (Sterck et al., 2012; Blanc-Mathieu et al., 2014). Using the tools available from the Pfam database (<http://pfam.xfam.org/>) (Punta et al., 2012), with default parameters (*E*-value 1.0), the protein domains in all protein sequences were identified. Potential homologous proteins between species were identified based on sequence similarity. We developed a variant of the Bidirectional Best Hit (BBH) that takes into account several candidates. We called this variant Multiple Bidirectional Best Hit (MBBH) and happen to be specially suited for duplication-enriched species such as *Arabidopsis*. The algorithm receives as input two fasta files containing the protein sequences of the two species to compare and the number *N* of multiple bidirectional best hits to consider (Figure 1). In this study *N* was fixed to 20. For each protein $Prot_i^1$ (encoded by $Gene_i^1$, deep green) from the first species, the *N* proteins (encoded by $Gene_{i,k}^2$, red) from the second species, $Prot_{i,1}^2, Prot_{i,2}^2, \dots, Prot_{i,N}^2$, exhibiting the highest sequence similarity are selected using the Needleman-Wunsch algorithm implemented in the R Bioconductor package Biostrings (Pagès et al., 2016). These proteins will be called “initial best hits.” The same process is carried out for species 2: For each protein $Prot_j^2$ (encoded by $Gene_{j,k}^2$, red) from the second species, the *N* proteins from the first species, $Prot_{j,1}^1, Prot_{j,2}^1, \dots, Prot_{j,N}^1$, exhibiting the highest sequence similarity are selected. Next, for each $Prot_i^1$ from the first species its initial best hits, $Prot_{i,1}^2, Prot_{i,2}^2, \dots, Prot_{i,N}^2$, are filtered. Only those $Prot_{i,k}^2$ that present $Prot_i^1$ in their *N* initial best hits are kept. The same process is carried out for each protein $Prot_j^2$ from species 2 keeping only the initial best hits that present $Prot_j^2$ among their *N* initial best hits. An additional filtering process is performed by removing from the best bidirectional hits those that do not share at least one domain with the corresponding protein. Finally, we also assign putative orthology to proteins showing no MBBH target, but sharing exactly identical number and order of one or more, non-overlapping pfam domains. This means that due to the specificity of pfam domains, the presence of a single domain does not immediately imply the assignation of a putative orthologue as would be the case with specific higher plant domains such as NAM (Figure 3C) or specific microalgal domains. In this way, our approach combines global sequence similarity with

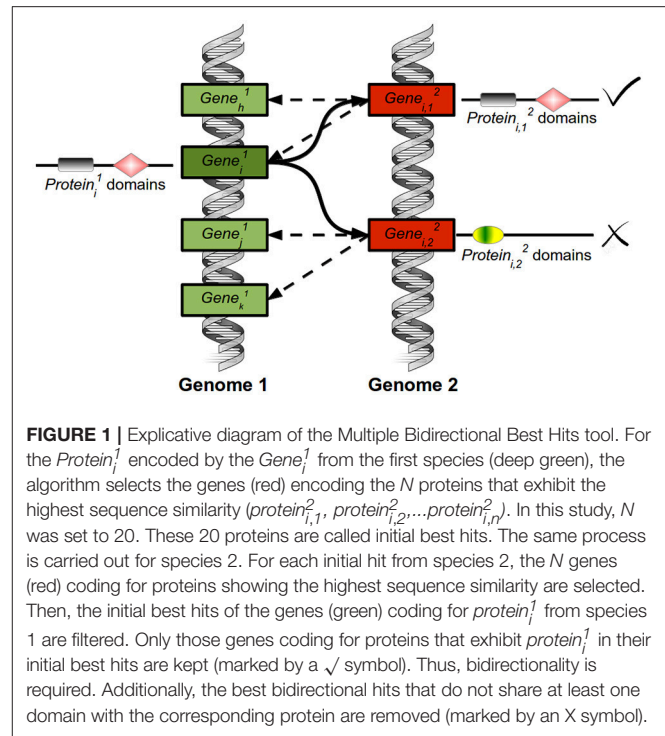


FIGURE 1 | Explicative diagram of the Multiple Bidirectional Best Hits tool. For the $Protein_i^1$ encoded by the $Gene_i^1$ from the first species (deep green), the algorithm selects the genes (red) encoding the *N* proteins that exhibit the highest sequence similarity ($protein_{i,1}^2, protein_{i,2}^2, \dots, protein_{i,N}^2$). In this study, *N* was set to 20. These 20 proteins are called initial best hits. The same process is carried out for species 2. For each initial hit from species 2, the *N* genes (red) coding for proteins showing the highest sequence similarity are selected. Then, the initial best hits of the genes (green) coding for $protein_i^1$ from species 1 are filtered. Only those genes coding for proteins that exhibit $protein_i^1$ in their initial best hits are kept (marked by a ✓ symbol). Thus, bidirectionality is required. Additionally, the best bidirectional hits that do not share at least one domain with the corresponding protein are removed (marked by an X symbol).

protein domain structure and, additionally, allows for multiple best hits in order to capture the evolution of a single gene into a multi-gene family. In general, MBBH could be a useful tool for those researchers trying to identify orthologues from any given gene in different species, even when they belong to distantly related ones. The tools developed in this article will be operative in the web site: (<http://viridiplantae.ibvf.csic.es/circadiaNet/>).

Transcriptomic Data Acquisition and Processing

For the three photosynthetic species analyzed in this study, transcriptomic data comprising time series of 48 or 72 h were collected from the GEO database (<https://www.ncbi.nlm.nih.gov/geo/>) (Barrett et al., 2013). In the case of *Arabidopsis*, two different data sets were used. The first one identified with the accession number GSE3416 (Bläsing et al., 2005) consists of microarray data taken in ND conditions (12:12). The second one identified with the accession number GSE43865 (Rugnone et al., 2013) consists of RNA-seq data taken in LD conditions (16:8). For *Chlamydomonas*, a single RNA-seq data set was collected identified with the accession number GSE71469 (Zones et al., 2015) taken in ND conditions. For *Ostreococcus*, a single microarray data set was used, identified with the accession number GSE16422 (Monnier et al., 2010) taken in ND conditions. The microarray data were processed using the Robust Multi-array Average (RMA) implemented in the Bioconductor R package affy (Gautier et al., 2004). The RNA-seq data was processed following the Tuxedo protocol (Trapnell et al., 2012). Reads were mapped to the reference genomes downloaded from Phytozome using Tophat. Transcripts were

TABLE 1 | Primers employed in QPCR experiments.

	Forward sequence	Reverse sequence	Amplified fragment size (bp)
<i>Arabidopsis</i> GENES			
<i>UBQ10</i>	5'-GAAGTTCATGTTTCGTTTCATGT-3'	5'-GGATTATACAAGGCCCAAAA-3'	145
<i>HY5</i>	5'-GCTGAAGAGGTTGTTGAGGA-3'	5'-TCTCCAAGTCTTTCACCTGTGTTT-3'	101
<i>Chlamydomonas</i> GENES			
<i>CrTUB</i>	5'-GTTGCATCGTTAGCGTGGACG-3'	5'-GCAGCAGCCAATGTTTCAGACT-3'	170
<i>CrHY5</i>	5'-GGCATGAACCCCTAGCTTCC-3'	5'-CATCATGTCCTCGCTGATGT-3'	144
<i>Ostreococcus</i> GENES			
<i>OtEF1α</i>	5'-GACGCGACGGTGGATCAA-3'	5'-CGACTGCCATCGTTTTACC-3'	202
<i>OtHY5</i>	5'-GAGGGAAGATTGGGAAGAAGAA-3'	5'-CCTTCTCCAACGCCTGAAT-3'	81

assembled using Cufflinks. Default parameters were used for the different software tools. Gene expression levels were measured as FPKM (Fragments Per Kilobase of exon per Million fragments mapped). Since RNA-seq and microarray gene expression levels have different measurement units these data were normalized by subtracting the mean and dividing by the standard deviation using the R function `scale` from the base package. This allows representing both types of data on a common scale. Samples were collected every 3 or 4 h in the data analyzed in this study. In order to produce a smooth representation of the gene expression profiles in heatmap graphs these data were linearly interpolated to produce time series consisting of 24 h using the R function `approx` from the stats package. In heatmaps, genes were sorted according to their expression profile similarity using hierarchical clustering.

Identification and Clustering of Rhythmic Daily Gene Expression Patterns

The detection of significant periodic patterns in the transcriptomic data analyzed was performed using RAIN (Rhythmicity Analysis Incorporating Nonparametric methods). This Bioconductor R package consists of functions that implement robust nonparametric methods for the detection of rhythms with arbitrary wave forms and pre-specified periods (Thaben and Westermark, 2014). The main function `rain` was used with the following parameters: A numeric matrix comprising the gene expression levels from the different biological replicates of the 24 h time series; a sampling interval of 4 h for *Arabidopsis* and 3 h for *Chlamydomonas* and *Ostreococcus*; a period of 24 h; and a number of three replicates for *Arabidopsis* and *Ostreococcus* and two for *Chlamydomonas*. A significance level $\alpha = 1\%$ was chosen to assume that a certain gene exhibits a rhythmic daily expression pattern. The wave form of a gene exhibiting rhythmic daily expression was characterized using its peak (time point where the highest expression value is reached) and its trough (time point where the lowest expression value is reached). Daily rhythmic genes were thus classified into 16 different clusters according to their peaks and troughs at a particular time of the day, that is, all genes in the same cluster present their peaks and troughs at the same time point.

Gene Ontology Term and Pathway Enrichment Analysis

Gene Ontology (GO) terms associated to each *Arabidopsis* and *Chlamydomonas* gene were downloaded from Phytozome. For *Ostreococcus*, GO terms were downloaded from ORCAE. The Bioconductor R package `topGO` (Alexa and Rahnenfuhrer, 2016) was used to determine GO terms significantly enriched in different gene sets. The entire genome of the corresponding species was used as gene background. The statistical significance test selected was Fisher's exact test with a significance level $\alpha = 5\%$. The web based tool REVIGO (Supek et al., 2011) was used to remove redundancy from the enriched GO terms and produced a summary. The identification of enriched pathways in *Arabidopsis* gene sets was performed with the Bioconductor R package `clusterProfiler` (Yu et al., 2012). The `enrichKEGG` function was used with Benjamin-Hochberg as p -value correction method for the multiple testing and a q -value cutoff of 0.05.

Gene Co-expression Network Construction, Visualization, and Analysis

Gene co-expression was measured using the correlation between gene expression profiles over the 24 h time series. Two daily rhythmic genes were assumed to be co-expressed when the Pearson correlation coefficient between their expression profiles was greater than 0.95. A gene co-expression network was constructed for each species where the nodes represent daily rhythmic genes and an edge is drawn between nodes when the corresponding genes are co-expressed according to the previous criterion. Cytoscape, a software tool for the representation and analysis of complex networks (Shannon et al., 2003), was used to visualize the gene co-expression networks applying the Prefuse Force-Directed Layout. The analysis of the networks was performed using the R package `igraph` (Csárdi and Nepusz, 2006). The scale-free property was tested using linear regression over the logarithmic transform of the degree distribution. The small-world property was tested by generating 10^4 random scale-free networks with the same number of nodes and edges as the corresponding network, using the `barabasi.game` function from the `igraph` package.

Module Conservation Analysis

Two daily rhythmic genes from two different species determined as potential orthologues using MBBH were defined as exhibiting a conserved daily pattern when both belonged to the same cluster (they presented their peak and trough in the same time interval) or when the Pearson correlation coefficient between their expression profiles were higher than 0.98. The conservation among the co-expression patterns of two different sets of genes from two different species were computed according to the *summary composite conservation statistic* “Zsummary” as defined in Langfelder et al. (2011). A Zsummary value lower than 2 indicates no conservation, a Zsummary value 2–10 implies moderate conservation, while a Zsummary greater than 10 constitutes evidence of a great level of conservation. The R package WGCNA (Langfelder and Horvath, 2012) and the function implemented therein were used.

Plant, Algal Material, and Growth Conditions

Three independent biological replicates for plants and algae were grown in a model SG-1400 phytotron (Radiber SA, Spain) under LD conditions with temperatures ranging from 22°C (day) to 18°C (night) and 75 $\mu\text{Em}^{-2}\text{s}^{-1}$ light intensity. *Arabidopsis thaliana* Col-0 wild type seeds were incubated 4 days at 4°C in the dark before sowing in MS plates. 12-day-old seedlings were collected every 4 h during a 24 h period. Time points were denoted as Zeitgeber Time *N* (ZTN) indicating the time point *N* hours after the lights are switched on in the phytotron (ZT0), mimicking dawn. *Chlamydomonas reinhardtii* wild type CW15 was grown in flasks with minimal Sueoka medium for 12 days. Similarly, *Ostreococcus tauri* wild type RCC 745 was grown in flasks with Keller medium (Keller et al., 1987) for 12 days. Algae were harvested every 4 h during a 24 h period.

RNA Isolation and QPCR

RNA was isolated from *Arabidopsis* seedlings (0.1 g leaf tissue), *Chlamydomonas* and *Ostreococcus* (pellet of 25 ml culture at exponential phase) using a modified TRIZOL (Invitrogen) protocol as described by the manufacturer. Briefly, the sample was mixed with 1 ml of TRIZOL and 0.2 ml of chloroform and the mixture was centrifuged at 16,000 g for 10 min at 4°C. The supernatant was treated with 1 volume of 100% (v/v) 2-propanol, incubated 15 min at room temperature and centrifuged at 16,000 g for 10 min at 4°C. The pellet was dissolved in 0.75 ml 3 M LiCl, incubated for $t > 10$ min at room temperature and centrifuged at 16,000 g for 10 min at 4°C. This pellet was washed with 80% (v/v) ethanol and centrifuged at 16,000 g for 10 min at 4°C. The final RNA sample was suspended in 21 μl of DEPC treated water and 1 μl quantified employing a ND-1000 Spectrophotometer (Nanodrop).

One micro gram of TRIZOL-isolated RNA was used to synthesize cDNA employing the Quantitect[®] Reverse kit (Qiagen) following the instructions recommended by the manufacturer. cDNA samples were diluted to a final concentration of 10 ng/ μL and stored at -20°C until QPCR was performed. Primers to amplify the 3' translated region of *AtHY5*, *CrHY5*, *OtHY5*, including *AtUBQ10*, *CrTUB*, and

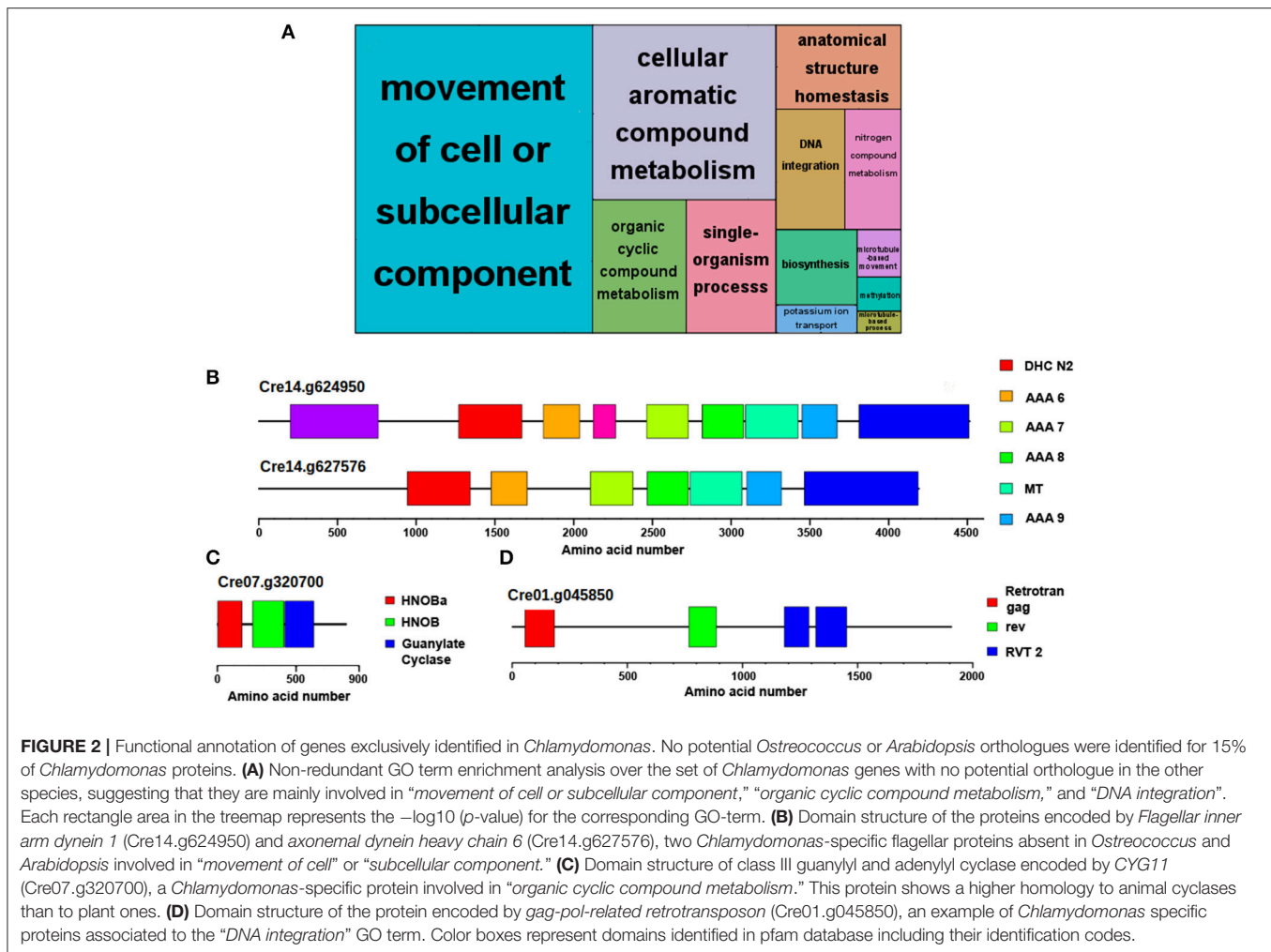
OtEF1 α as housekeeping genes (Table 1) were designed using the Oligo analyzer program from Integrated DNA Technologies (<http://eu.idtdna.com/analyzer/Applications/OligoAnalyzer/>). QPCR was performed in a Multicolor Real-Time PCR Detection System iQTM5 (Bio-Rad) in a 10 μL reaction: primer concentration 0.2 μM , 10 ng cDNA and 5 μL SensiFAST TM SYBR & Fluorescein Kit (Bioline). The QPCR program consisted in (i) 1 cycle (95°C, 2 min); (ii) 40 cycle (95°C, 5 s; 60°C, 10 s and 72°C, 6 s) (iii) 1 cycle (72°C, 6 s). Fluorescence was measured at the end of each extension step and the melting curve was performed between 55 and 95°C. Three biological replicates with three technical replicates from each species were used for every time point. The QPCR results were estimated using the ddCt R Bioconductor package (Zhang et al., 2015).

RESULTS AND DISCUSSIONS

Most *Ostreococcus* Proteins Present Potential Orthologues in *Arabidopsis* and *Chlamydomonas*, but This Is Not the Case with the Other Two Species

The identification of potential orthologues between two species constitutes one of the most important bottlenecks in comparative genomics and transcriptomics (Dessimoz et al., 2012). The Bidirectional Best Hit (BBH) algorithm is the most commonly used method for the automated identification of putative orthologues. In spite of its simplicity, BBH is highly accurate when dealing with bacterial and archaeal genomes (Wolf and Koonin, 2012). Nevertheless, BBH does not perform optimally, missing as much as 60% of orthologues, in gene-duplication-enriched genomes such as those of plants and animals (Dalquen and Dessimoz, 2013). In this work, in order to identify orthology among distantly related species, we have developed a variant of the BBH algorithm termed Multiple Bidirectional Best Hit (MBBH). MBBH assigns to each gene g_i^l from the first species $k \leq N$ potential orthologues from the second species, g_1^k, \dots, g_k^k , based on sequence similarity of the proteins they encode if, and only if, the query gene g_i^l is among the N most similar genes from the first species (Section Materials and Methods, Figure 1).

Using MBBH tool, more than 97% of *Ostreococcus* proteins could be ascribed to a potential orthologue either in *Arabidopsis* or *Chlamydomonas* (Supplementary Table 1). This is in agreement with the reduced and compact *Ostreococcus* genome (Palenik et al., 2007) and suggests, as has been proposed (Raven et al., 2013), that this marine picoeukaryote contains the minimal genome for a functional photosynthetic eukaryote and that it is almost entirely regulated by the circadian clock (see below). Confirming this idea, a GO term enrichment analysis over the set of genes without potential *Arabidopsis* or *Chlamydomonas* orthologues did not produce any significant result, indicating that these genes are not involved in any specific biological process. In fact, the highest number of genes without potential orthologues was located on chromosomes 2 and 19. Some of these genes codified for a number of unnamed products (ostta02g03690, ostta02g05500, ostta02g00800) and a group of Tc1-like/mariner transposases (ostta02g01245, ostta02g01247, ostta02g02355).



This supports previous analysis on the heterogeneity of *Ostreococcus* genome that identified chromosomes 2 and 19 as different from the other chromosomes in terms of GC content, codon usage and number of transposable elements (Derelle et al., 2006).

Following the same comparative analysis, approximately 85% of *Chlamydomonas* proteins present potential orthologues in *Ostreococcus* and *Arabidopsis* (Supplementary Table 2). Therefore, a set of approximately 2,600 *Chlamydomonas* genes that codify for specific proteins could not be traced to any orthologue in the other two species. An ontology enrichment analysis performed over this set of genes using the R Bioconductor package topGO and summarized with REVIGO revealed that the top three most significant and non-redundant GO terms were “Movement of cell or subcellular component,” “Organic cyclic compound metabolism,” and “DNA integration” (Figure 2A). The same set of genes were associated with the GO terms “Movement of cell or subcellular component,” “Anatomical structure homeostasis” and “single-organism process.” Therefore, the last two GO terms were considered redundant and only “Movement of cell or subcellular component”

is discussed. Similarly, the GO terms “Organic cyclic compound metabolism” and “cellular aromatic compound metabolism” included the same list of associated genes than “Organic cyclic compound metabolism,” so only the latter is discussed. Genes codifying flagellar proteins such as *flagellar inner arm dynein 1* (Cre14.g624950) and *axonemal dynein heavy chain 6* (Cre14.g627576) (Figure 2B) are representative of the group “Movement of cell or subcellular component.” *Chlamydomonas* cells exhibit two polar flagella whereas no flagellar-like structures are present in *Ostreococcus* or *Arabidopsis* cells, explaining the lack of orthologues in the non-flagellated organisms. Besides, the *Chlamydomonas* specific genes annotated with the GO term “Organic cyclic compound metabolism” include the class III guanylyl and adenylyl cyclase family including genes such as *adelynate/guanylate cyclase CYG11* (Cre07.g320700) (Figure 2C). These enzymes that catalyze the synthesis of cGMP and cAMP in animals are absent in plants, hence the lack of potential orthologues in *Ostreococcus* and *Arabidopsis*. In fact, this concurs with the idea that some *Chlamydomonas* genes are closer to animal than to plant ones (Merchant et al., 2007). On the other hand, the unique *gag-pol-related*

retrotransposon (Cre01g.045850) (Figure 2D) is an example of the *Chlamydomonas* specific genes associated with “DNA integration” GO term, suggesting that this set of viral pathogens infecting *Chlamydomonas* is specific for the alga.

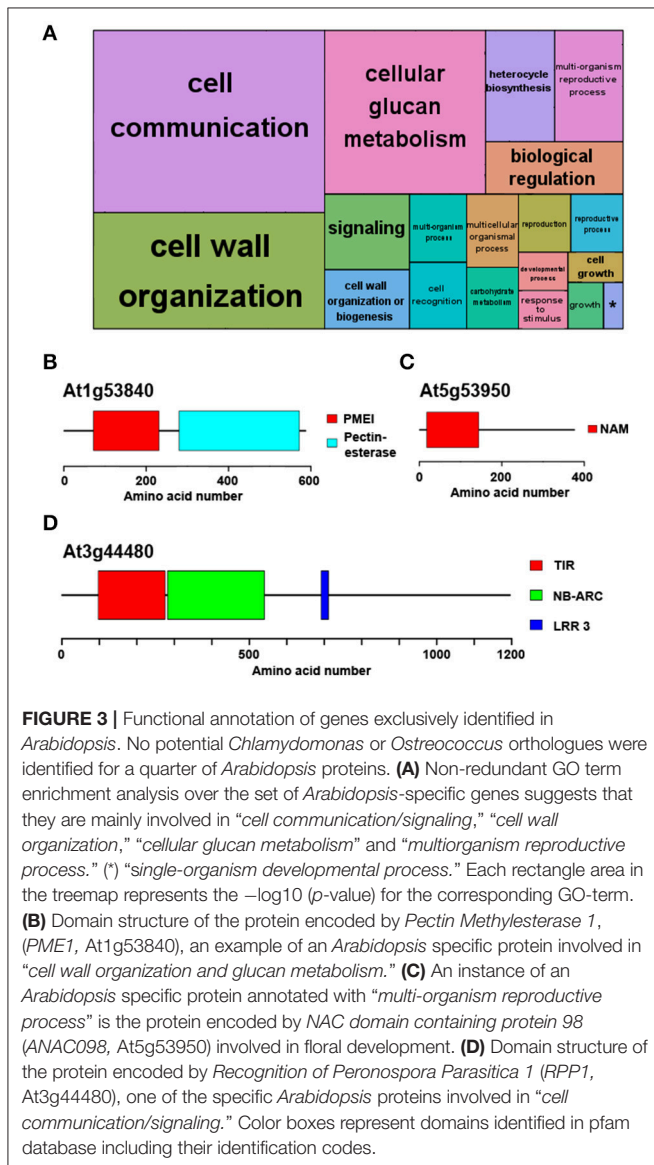
The analysis also identified potential *Ostreococcus* and *Chlamydomonas* orthologues for approximately 75% of *Arabidopsis* proteins (Supplementary Table 3), suggesting that one fourth of the *Arabidopsis* proteins are not present in microalgae and have potentially been acquired during the course of higher plant evolution. In accordance with this idea, a GO term enrichment analysis showed that these *Arabidopsis*-specific proteins are involved in the following top three most significant non-redundant terms: “Cell communication,” “Cell wall organization,” and “Multiorganism reproductive process” (Figure 3A). Similar to the previous result, the GO term “Signaling” was found to share the same gene list with “Cell

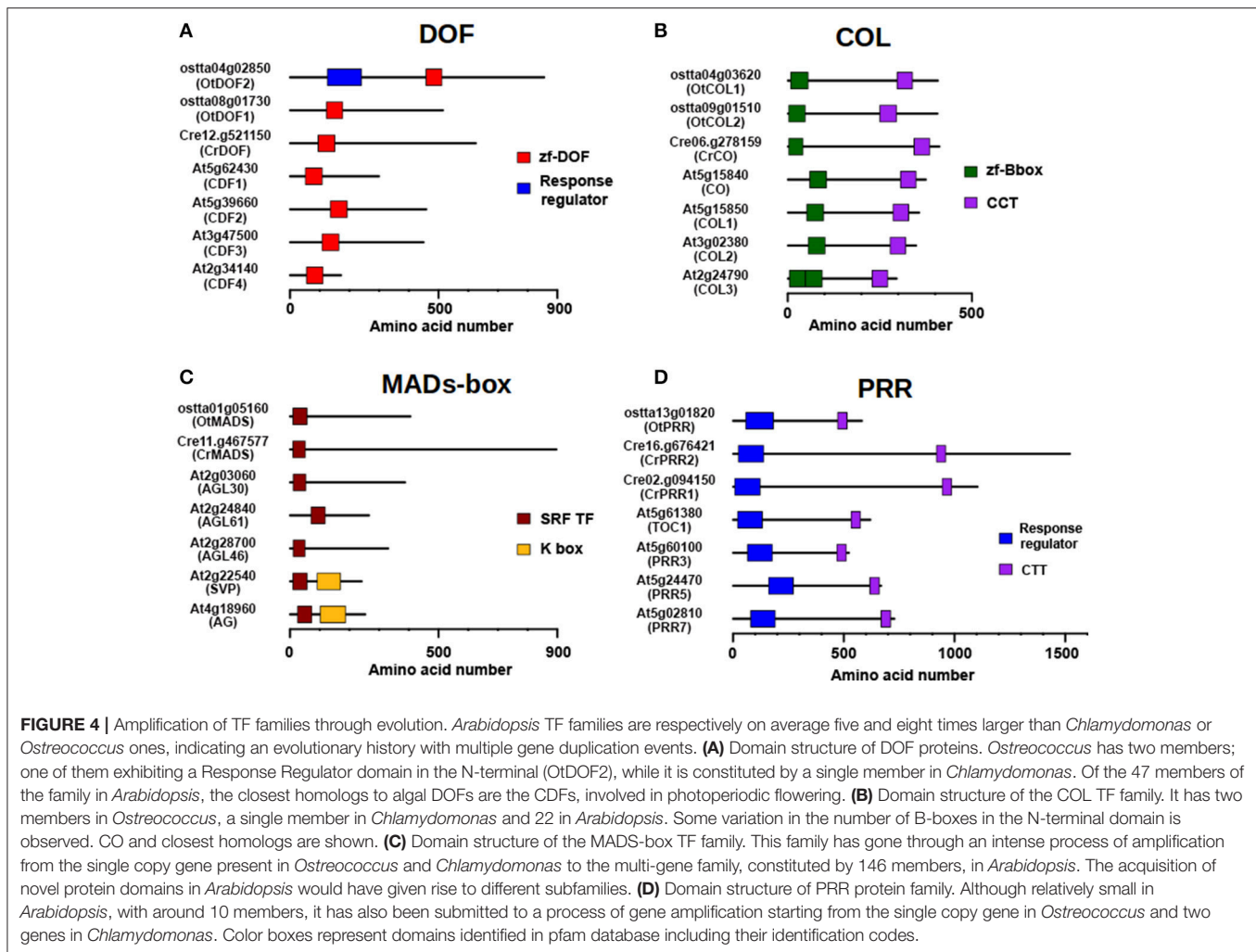
communication”; “Cellular glucan metabolism” presented the same associated genes as “Cell wall organization” and the set of genes assigned to “Biological regulation” and “Multiorganism reproductive process” were identical. Therefore, the former GO terms were considered redundant and only the latter GO terms are discussed. These GO terms correspond to complex multicellular plants features that are absent in microalgae. This way, key proteins involved in cell wall biogenesis and cellular glucan metabolism in *Arabidopsis* such as *Pectin Methylesterase 1* (*PME1*, At1g53840) lack *Ostreococcus* and *Chlamydomonas* potential orthologues (Figure 3B). Indeed, the evolutionary history of *PME* genes has been previously studied, establishing their appearance in multicellular Charophyte algae and supporting their absence in more primitive algae (Wang et al., 2013). A case of proteins annotated as involved in “Multi-cellular processes” without potential orthologues in *Ostreococcus* and *Chlamydomonas* is *Arabidopsis* NAC domain containing protein 98 (*ANAC098*, At5g53950) (Figure 3C). The family of NAC TFs is one of the largest in plants and is involved in multiple key developmental processes such as floral development. Again, this TF family first originated in Charophytes (Zhu et al., 2012) and is absent in *Chlamydomonas* and *Ostreococcus*. Finally, the *Recognition of Peronospora Parasitica 1* (*RPP1*, At3g44480) is an example of a specific *Arabidopsis* protein involved in “Cell communication/signaling” and is part of the set of specific *Arabidopsis* genes related to pathogen resistance and programmed cell death (Schreiber et al., 2016; Figure 3D).

Large TF Families in *Arabidopsis* Have Single Orthologues in Microalgae Suggesting Gene Amplification and Functional Diversification Processes

The approach to define potential orthologues by MBBH can detect several candidates for any given gene, identifying multiple orthologous genes that could have appeared from processes of gene duplication. In *Arabidopsis*, on average, 5.19 and 7.83 genes could be ascribed to a homolog in *Chlamydomonas* and *Ostreococcus*, indicating that *Arabidopsis* gene families are, on average, five to eight times larger than in those organisms. This concurs with the idea that whole genome duplication and gene duplication events were crucial in the evolution of plant gene families (Romero-Campero et al., 2013; Rensing, 2014). Interestingly, this process is particularly frequent in TF families and MBBH tool has efficiently detected functional domains in the protein sequences of the three species under study and identified the TF family they belong to. In general, their sizes coincided with the data available in the Plant Transcription Factor Database, PlantTFDB (Jin et al., 2017), confirming the accuracy of this approach. On average, less than 4 and 7 protein members formed each *Ostreococcus* and *Chlamydomonas* TF family respectively, whereas a media of 30 members constituted *Arabidopsis* TF families, further supporting the idea of multiple duplication events.

Two clear examples of amplification in TFs are the DOF and COL protein families (Figures 4A,B) that present one single member in *Chlamydomonas* and two members in *Ostreococcus*,





respectively, whereas in *Arabidopsis* 47 DOF and 22 COL proteins are present. A subset of plant DOF genes are regulated by the clock (*CYCLING DOF FACTORS*, CDFs) and, in time, control the daily expression of *CONSTANS* (CO) in *Arabidopsis* and potato during photoperiodic flowering (Imaizumi et al., 2005; Fornara et al., 2009; Kloosterman et al., 2013). This constitutes a conserved DOF-CO module in Spermatophytes that is also conserved in *Chlamydomonas* (Lucas-Reina et al., 2015). In *Ostreococcus* it might be different. While the *Chlamydomonas* and *Arabidopsis* orthologues presented a single DOF domain, one of the two *Ostreococcus* DOF proteins (OtDOF2, ostta04g02850) presented an additional N-terminal Response Regulator domain with a potential phosphoacceptor aspartic acid-aspartic acid-lysine (DDK) motif (Figure 4A). This suggests that OtDOF2 could be part of a phosphorelay system (Djouani-Tahri et al., 2011). The COL gene family (Figure 4B) seemed to have appeared in microalgae, with a single representative in *Chlamydomonas* (CrCO, Cre06.g278159) (Valverde, 2011) and two putative orthologues in *Ostreococcus* (OtCOL1, ostta04g03620 and OtCOL2, ostta09g01510). In *Chlamydomonas* CrCO is involved in the control of the cell cycle, starch synthesis and oil content (Serrano et al., 2009; Deng

et al., 2015) and some of these functions have been conserved in some *Arabidopsis* orthologues (Romero-Campero et al., 2013; Ortiz-Marchena et al., 2014).

The MADS-box and PRR families (Figures 4C,D) constitute other interesting cases. MADS-box TF family is an example of frequent amplification over the course of plant evolution, since only a single copy is present in *Chlamydomonas* and *Ostreococcus*, whereas in *Arabidopsis* this family contains 146 members. MADS-boxes have been classified into different subfamilies depending on the recruitment of new protein domains besides the conserved MADS DNA-binding domain, such as the K-boxes (Figure 4C), and that may explain the ample functional diversity of these TF family in higher plants. On the other hand, modern plant PRRs (Figure 4D) contain an N-terminal Pseudo Response Regulator domain and a C-terminal CCT domain, but algal proteins are different. Spermatophytic PRRs are similar in size to COLs and, like them, seem to have experienced a similar amplification process from a single gene copy in *Ostreococcus* (OtPRR, ostta13g01820) and two genes in *Chlamydomonas* (CrPPR1, Cre02g.094150; CrPPR2, Cre16.g676421) to a multi-gene family in *Arabidopsis*. Nevertheless, while in *Ostreococcus* and *Chlamydomonas*

proteins, a potential true phosphoacceptor DDK motif is present, suggesting that the algal proteins still retain part of the ancestral phosphorelay signaling mechanism, this motif is missing in the higher plant domain, constituting a “Pseudo” Response Regulator (Mizuno and Nakamichi, 2005; Satbhai et al., 2011).

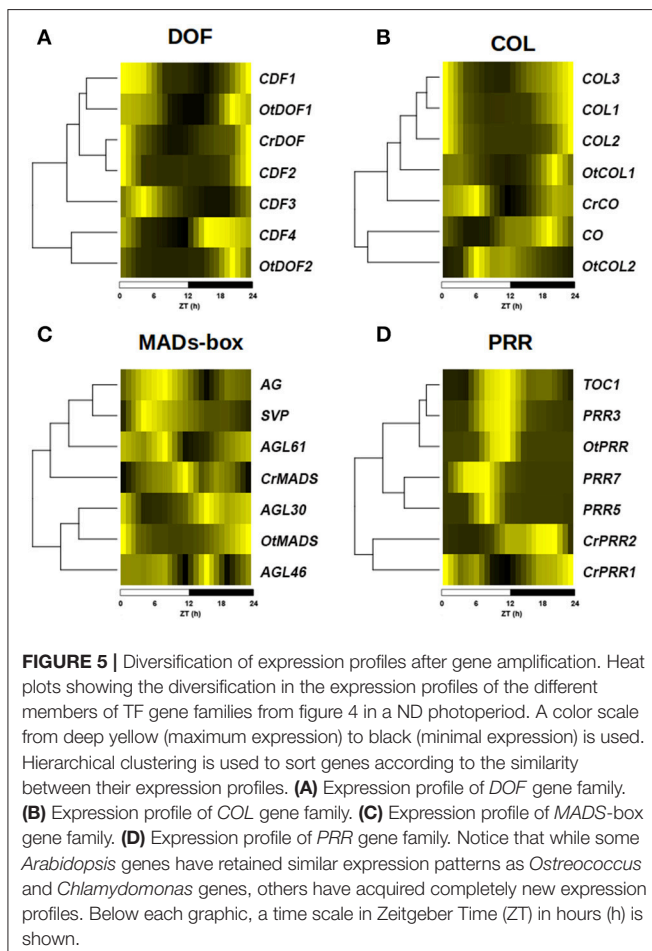
To study to what extent the conservation in protein sequence and domain structure is accompanied by conservation in the expression profiles, transcriptomic data consisting of 24 h time series for the three species under study was analyzed. Not surprisingly, differences in the expression profiles of *Arabidopsis* genes were observed, some genes retaining the same one as in *Ostreococcus* and *Chlamydomonas*, while others acquired completely new expression profiles (Figure 5). This way, in the *DOF* family (Figure 5A) the expression profile of the *Arabidopsis* genes *CDF1* and *CDF2* have retained similar expression profiles as *CrDOF* and *OtDOF1* with a peak around ZT0 and a trough around ZT12. However, *CDF4* and *OtDOF2* peak around ZT21, while *CDF3* exhibits an expression profile with a peak at ZT3 and trough at ZT18, not observed in *Chlamydomonas* or *Ostreococcus* genes. A similar situation is observed for the *COL* (Figure 5B), *MADS*-box (Figure 5C) and *PRR* (Figure 5D) TF families. The *Arabidopsis* *COL* genes *COL1*, *COL2* and *COL3* exhibit very similar expression profiles to *OtCOL1* and *CrCO*, showing

a trough at ZT12 (Figure 5B). On the other hand, *OtCO2* and *CO* present distinctly different expression profiles with their minimum expression levels at ZT0 and ZT6, respectively. The *MADS*-box family presents a great diversification in the expression profile of their members, with a substantial difference between the expression patterns of *Arabidopsis*, *Chlamydomonas* and *Ostreococcus* genes (Figure 5C). In the *PRR* family, *OtPRR* presents the same expression pattern as *Arabidopsis* *PRR1* (*TOC1*) and *PRR3*, peaking at ZT12, whereas *PRR5* and *PRR7* peak earlier, at around ZT9. However, *CrPRR1* and *CrPRR2* present completely different profiles, peaking at ZT21 and ZT0, respectively (Figure 5D). This could imply that, in general, gene amplification is accompanied by expression profile diversification and hence, functional diversification (Romero-Campero et al., 2013).

Most Genes in *Chlamydomonas* and *Ostreococcus* Exhibit Light-Dependent Daily Rhythmic Expression Patterns, but Only 40% of the *Arabidopsis* Transcriptome Shows a Periodic Expression

Recently, massive amounts of transcriptomic data for photosynthetic organisms under diverse environmental and physiological conditions have been produced. One of the most studied environmental signals is the alternation of light and dark cycles. Nevertheless, the analysis of the diurnal changes in the transcriptome has been independently performed in different species, making necessary the application of Molecular Systems Biology techniques to integrate and compare them. In this study we have analyzed microarray data collected over 24 h periods in ND conditions for *Ostreococcus* (Monnier et al., 2010) and *Arabidopsis* (Bläsing et al., 2005), as well as RNA-seq data for *Chlamydomonas* (Zones et al., 2015) in ND conditions and *Arabidopsis* in LD conditions (Rugnone et al., 2013). Details on transcriptomic data processing are described in the Materials and Methods section. In order to determine the effect of the alternation of light and dark periods over the transcriptome we identified genes exhibiting rhythmic or oscillating expression patterns with a period of approximately 24 h in the three species under study. Nonparametric methods implemented in the Bioconductor R package RAIN (Rhythmicity Analysis Incorporating Nonparametric methods) were used to detect significant gene expression patterns with arbitrary wave forms and a pre-specified period of 24 h (Thaben and Westermark, 2014), see Materials and Methods section for details.

The analysis revealed that more than 90% of the *Ostreococcus* genes targeted in the microarray exhibited daily rhythmic patterns (Supplementary Table 1). No GO term significantly enriched in the non-periodic genes was identified. This suggests that practically the entire *Ostreococcus* transcriptome is periodic and that most *Ostreococcus* biological processes are strongly affected by the alternation of light and dark cycles. In *Chlamydomonas* approximately 70% of the genes were found to follow significant daily rhythmic expression patterns (Supplementary Table 2). The GO term enrichment analysis over the non-periodic *Chlamydomonas* genes identified only a



few significant biological processes. The two most significant processes were “DNA integration” and “Defense response to virus” including genes such as the reverse transcriptase Cre05.g235102, the DNA ligase Cre06.g277801 and the 2'-5' oligoadenylate synthetase Cre15.g641050. These have been identified as biological processes induced by biotic stimuli, and are thus independent from the abiotic inputs from light/dark cycles. Finally, only 43.18% of the *Arabidopsis* genes showed significant circadian patterns according to our analysis (Supplementary Table 3), suggesting that during the evolution of higher plants many biological processes were uncoupled from the external influence of alternating light and dark cycles. These percentages of rhythmic genes are largely in agreement with previous results (Bläsing et al., 2005; Monnier et al., 2010; Zones et al., 2015). The pathway enrichment analysis performed over the *Arabidopsis* daily rhythmic genes revealed several key significant pathways influenced by rhythmic changes (Table 2). As expected, the most significant pathway represented in the enrichment was the one termed “Circadian rhythms in plants.” Also expected were the pathways “Porphyrin and chlorophyll metabolism” and “Pentose phosphate pathway”. These pathways are involved in photosynthesis and hence are expected to be highly regulated by light/dark cycles exhibiting rhythmic patterns. Key metabolic pathways such as “Fatty acid degradation” and “Starch and sucrose metabolism” were also detected as significantly enriched in daily rhythmic genes indicating that *Arabidopsis* metabolism is still highly affected by alternating periods of light and dark following rhythmic patterns. Somehow surprising was the identification of the “Alpha-Linolenic acid metabolism” and “Plant hormone signal transduction” among the enriched GO groups. This result suggests either that these pathways are ancient or that during the course of plant speciation new or exclusive plant pathways involved in hormone synthesis and sensing were recruited and acquired a daily rhythmic regulatory pattern. Curiously, although the course of evolution seems to have made the clock more complex in *Arabidopsis*, a lower percentage of the transcriptome seems to be regulated in a light-dependent periodic manner. It could be that a precise, fine tuning of the clock was recruited to make new processes independent from light/dark transitions in higher plants. More circadian experiments in different tissues or developmental stages may be needed to explain this apparent paradox.

With the aim of representing the daily rhythmic genes and their complex co-expression relationships, a gene co-expression network for each photosynthetic species analyzed in this study was constructed (Figures 6A–C). Nodes represent daily rhythmic genes and edges co-expression relationships, so that two circadian genes are assumed to be co-expressed when the Pearson correlation index between their 24 h expression profiles is greater than 0.95 and an edge is drawn between them. The three gene co-expression networks were visualized using the Prefuse Force Directed layout implemented in the software tool Cytoscape (Shannon et al., 2003). Interestingly, the three different gene co-expression networks acquired the same ring-like structure, capturing the chronological relationship between the co-expression patterns of the periodic genes. The basic topological parameters in network analysis, namely node

TABLE 2 | Significantly enriched pathways among the *Arabidopsis* circadian genes.

Pathway ID	Description	q-value	Ratio circadian/total
ath04712	Circadian rhythm-plant	$4.20 \cdot 10^{-6}$	29/36
ath00030	Pentose phosphate pathway	$5.25 \cdot 10^{-3}$	35/58
ath00052	Galactose metabolism	$1.54 \cdot 10^{-2}$	32/55
ath00590	Arachidonic acid metabolism	$1.54 \cdot 10^{-2}$	13/17
ath00966	Glucosinolate biosynthesis	$1.54 \cdot 10^{-2}$	14/19
ath00860	Porphyrin and chlorophyll metabolism	$1.76 \cdot 10^{-2}$	28/48
ath00360	Phenylalanine metabolism	$1.84 \cdot 10^{-2}$	25/42
ath00071	Fatty acid degradation	$2.66 \cdot 10^{-2}$	24/41
ath00500	Starch and sucrose metabolism	$2.66 \cdot 10^{-2}$	66/139
ath00130	Ubiquinone and other terpenoid-quinone biosynthesis	$2.66 \cdot 10^{-2}$	21/35
ath00260	Glycine, serine and threonine metabolism	$3.54 \cdot 10^{-2}$	37/72
ath00592	alpha-Linolenic acid metabolism	$3.54 \cdot 10^{-2}$	21/36
ath01200	Carbon metabolism	$3.62 \cdot 10^{-2}$	114/262
ath01230	Biosynthesis of amino acids	$3.62 \cdot 10^{-2}$	111/255
ath04075	Plant hormone signal transduction	$3.62 \cdot 10^{-2}$	118/273
ath01210	2-Oxocarboxylic acid metabolism	$4.66 \cdot 10^{-2}$	37/74

degree distribution and clustering coefficient, were computed for each network (Figure 6D). The degree distribution of the three networks follows an exponential negative distribution with p -values below $2.2 \cdot 10^{-16}$, which implies that they are scale-free networks (Barabási, 2015), suggesting that the global daily rhythmic co-expression patterns in the three species are robust to random changes or mutations but fragile to direct changes or mutations in hub genes, those co-expressed with a high number of genes (Aoki et al., 2007). Additionally, the clustering coefficient of the three networks was significantly high when compared to the clustering coefficient of random scale-free networks with the same number of nodes and edges. Therefore, all three networks constitute small-world networks (Barabási, 2015) with expected short paths connecting any two nodes of the network. This is assumed to facilitate the quick propagation of information between genes with different periodic expression patterns.

Daily Rhythmic Genes Clusters Constitute Transcriptional Programs That Confer Temporal Separation to Different Biological Processes with Different Levels of Conservation between *Ostreococcus*, *Chlamydomonas*, and *Arabidopsis*

The daily rhythmic expression pattern of individual genes can be described by using its peak, the time point when the expression

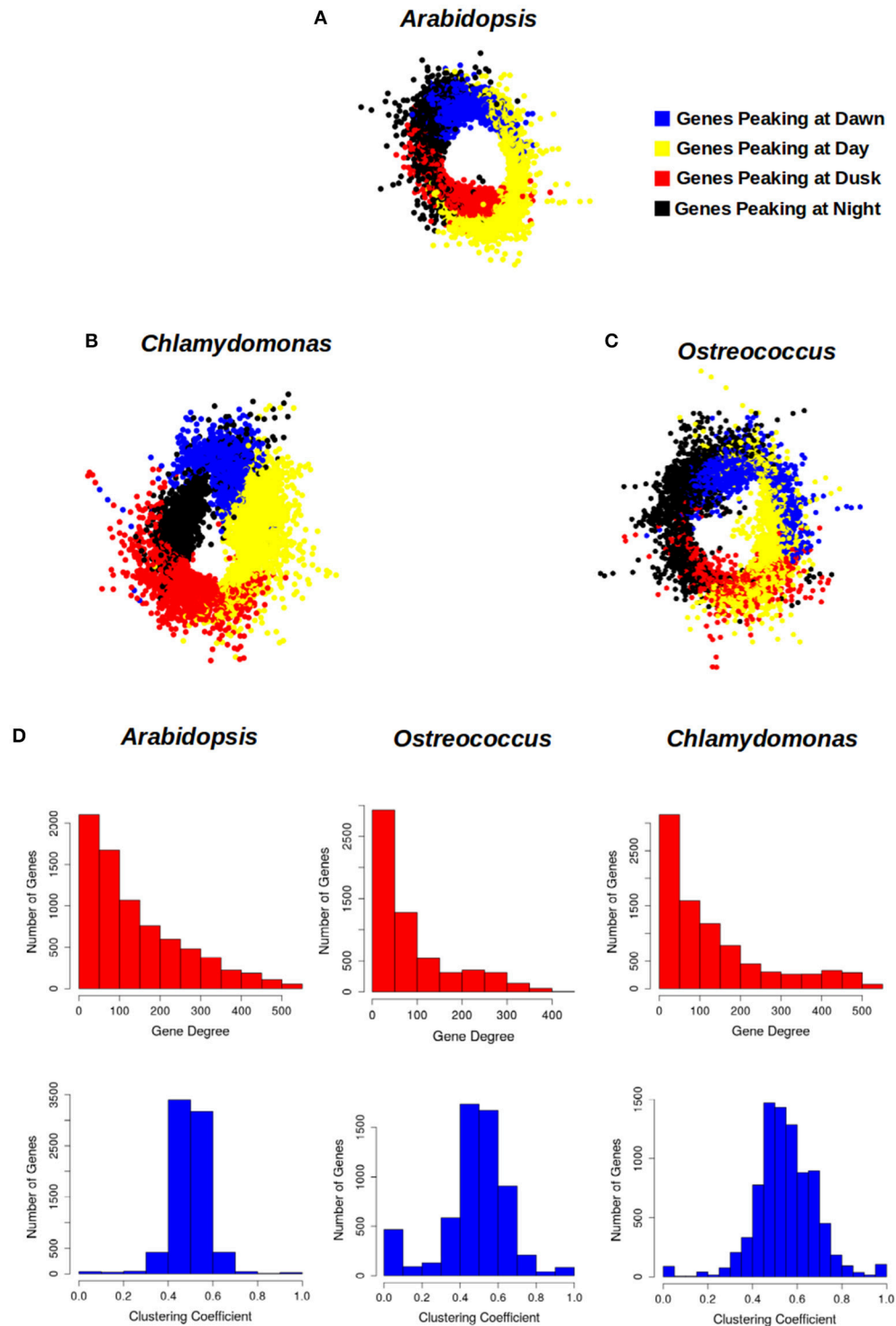


FIGURE 6 | Circadian gene co-expression networks. Genes are represented by nodes and edges represent co-expression between them (Pearson correlation index between their 24 h-expression profiles greater than 0.95). The three networks acquired a ring-like structure mirroring the periodic expression pattern of daily rhythmic genes. **(A)** *Arabidopsis* daily rhythmic gene co-expression network comprised 7639 circadian genes with 537027 co-expression relationships. **(B)** *Chlamydomonas* daily rhythmic gene co-expression network consisted of 10338 circadian genes and 573980 co-expression interactions. **(C)** *Ostreococcus* daily rhythmic gene co-expression network comprised 5782 nodes and 222493 co-expression relationships. **(D)** The topological analysis of the three gene co-expression network shows that they are scale-free and small-world networks: (1) Their degree distributions (above, red) follow an exponential negative distribution. (2) Their average clustering coefficient (below, blue) is significantly greater than random networks with equal number of nodes and edges. Gene cluster with a peak at Dawn is represented in blue, that with a peak during the Day in yellow, the one showing a peak at Dusk in red and that with a peak at Night in black.

to be from ZT15 to ZT21 (**Figure 7B**). Thus, daily rhythmic genes could be classified into 16 different clusters according to the time interval where their peak was located (Dawn, Day, Dusk, and Night) and the time interval containing their trough (again Dawn, Day, Dusk, and Night). Due to the amount of data engaged, the graphic representation of these 16 clusters can be confusing, so, in order to facilitate their visualization, we decided to merge all clusters peaking at the same interval into a single one. This way, for each generalized gene cluster we represented the expression profile of every gene and the average expression profile of the entire cluster (**Figure 7C**). Interestingly, no major differences were apparent between the average expression profiles from the three different species under analysis. Furthermore, when we colored the location of these clusters in the corresponding gene co-expression networks (**Figures 6A–C**), we observed that these clusters presented the same localization in the three different networks. This is a strong support for the veracity of our approach and shows, at a single view, that daily rhythmic genes associate in true temporal sets of genes performing synchronous actions and roughly reflecting the natural course of a 24 h clock. In this analysis the boundaries separating clusters were more sharply defined in *Arabidopsis* and *Chlamydomonas* than *Ostreococcus* that showed a higher degree of genes mixing between temporal clusters (**Figures 6A–C**). This supports the already established idea that the primitive *Ostreococcus* clock is not as efficient as those of the other more complex organisms.

The two largest gene clusters in *Ostreococcus* and *Chlamydomonas* contain genes peaking at Dusk with their troughs at Dawn (1620 and 3751 genes, respectively) and genes peaking at Dawn with their troughs at Dusk (1744 and 2869 genes, respectively; **Table 3**). The genes of these two clusters comprise 56.85 and 49.98% of the entire set of daily rhythmic genes in *Ostreococcus* and *Chlamydomonas*, respectively and concur with the common notion that the light/dark and dark/light transitions play central roles in daily rhythmic gene expression regulation in both algae. Surprisingly, in *Arabidopsis* these two gene clusters are not the largest ones, comprising only 22.51% of all periodic genes. This suggests a degree of uncoupling between regulation of circadian gene expression and light/dark and dark/light transitions in higher plants. Nevertheless, the largest gene cluster in *Arabidopsis* is made up of 1209 genes peaking at Dawn with their troughs in the Day, indicating that the dark/light transition still plays a key role in the regulation of daily rhythmic genes in *Arabidopsis*. The second largest cluster in *Arabidopsis* contains 1014 genes whose expression peaks take place at Night and troughs during the Day. This suggests certain independence in the regulation of daily rhythmic gene expression in *Arabidopsis* from the light/dark transitions since a large number of genes present their peaks or troughs in the periods of light and dark.

We performed GO term enrichment analysis over the different gene clusters (**Table 3**) in order to determine the biological processes that were carried out by the different rhythmic gene clusters. Different clusters were enriched in distinct biological processes indicating a temporal separation among them. Some biological processes show enrichment in the same clusters for

the three different species such as “DNA metabolic process” (gene clusters with peak at Dusk and trough at Dawn), “photosynthesis” (gene clusters with peak at Day and trough at Night) and “carbohydrate catabolic process” (gene clusters that have the peak during the Dawn and trough at Dusk) indicating a conservation in their daily rhythmic expression patterns (**Table 2**). On the contrary, an anticipation, delay or uncoupling from periodic expression of certain biological processes was apparent when the three species were compared.

Cell cycle-related GO terms such as “DNA metabolic process” were enriched in the gene cluster with peak at Dusk and trough at Dawn in the three species, including genes such as *REPLICATION PROTEIN A* (Cre16.g65100, ostta18g01440 and At4g19130), *ORIGIN RECOGNITION COMPLEX 1* (Cre10.g455600, ostta04g05220, and At1g26840) and *CELL DIVISION CYCLE 45* (Cre06.g270250, ostta04g04640, and At3g25100) that showed very similar expression profiles in the three species (**Figure 8A**). Interestingly, the GO terms “DNA replication” and “DNA metabolic process” are also enriched in the gene cluster with peak at Day and trough at Dawn only in *Ostreococcus* and *Arabidopsis* indicating an apparent anticipation or broader gene expression peaks in these two species with respect to *Chlamydomonas*, where narrow peaks are observed at the light/dark transition (**Figure 8A**). For instance, *DNA POLYMERASE A4* (Cre07.g312350, ostta13g02040, and At5g41880), *MINI CHROMOSOME MAINTENANCE 2* (Cre07.g338000, ostta01g02580, and At1g44900), *CYCLIN A1* (Cre03.g207900, ostta02g00150, and At1g44110) present a narrow peak centered on ZT12 in *Chlamydomonas*, whereas in *Ostreococcus* and *Arabidopsis* they present a broader peak centered on ZT6. This could also indicate a better culture synchronization in *Chlamydomonas* than *Ostreococcus* and the lack of synchronicity between the different tissues in *Arabidopsis*. The GO term “Photosynthesis” appeared significantly enriched in the gene clusters with peak at the Day and trough at Night in the three different species (**Table 3**). Specifically, *PHOTOSYSTEM I SUBUNIT D* (Cre05.g238332, ostta10g03280, and At1g03130) and *PHOTOSYSTEM II SUBUNIT O* (Cre09.g396213, ostta14g00150, and At3g50820) exhibit very similar expression patterns with broader peaks in *Ostreococcus* and *Arabidopsis* than in *Chlamydomonas* (**Figure 8B**). Only small variations were detected, for instance, in the expression profiles of genes codifying for components of plastid ATP synthase and b6f complex. *ATP CHLOROPLAST SYNTHASE 1* (*ATPC*, Cre06.g259900, ostta09g01080, and At4g04640) and *PHOTOSYNTHETIC ELECTRON TRANSFER C* (*PETC*, Cre11g.467689, ostta07g02450, and At4g03280) present a peak at dawn and trough at dusk in *Chlamydomonas* and *Ostreococcus*, while in *Arabidopsis* *ATPC1* is delayed, peaking during the day and *PETC* anticipates dawn, peaking at night (**Figure 8B**).

Ribogenesis, the process of ribosome making, is another interesting example of conservation and evolution of the periodic pattern of a biological process. In fact, “Ribosome biogenesis” seems to be enriched mainly in the gene cluster with peak at Dawn and a trough during the Day in *Ostreococcus* and *Chlamydomonas*, whereas in *Arabidopsis* is

TABLE 3 | GO term enrichment in the circadian gene clusters.

	<i>Arabidopsis thaliana</i>			<i>Ostreococcus tauri</i>			<i>Chlamydomonas reinhardtii</i>		
	Size	Enriched GO terms	Genes	Size	Enriched GO terms	Genes	Size	Enriched GO terms	Genes
Peak dawn	0			422	Protein catabolic process	<i>ostta1g00020</i>	1,426	Carboxylic acid metabolic process	<i>Cre16.g687350</i>
Trough dawn					Transcription from RNA polymerase II promoter	<i>ostta14g01300</i>		Cellular amino acid metabolic process	<i>Cre01.g029250</i>
Peak dawn	1,209	Carbohydrate catabolic process	<i>AT4G17090</i>	1,384	RNA metabolic process	<i>ostta10g00100</i>	1,801	Ribosome biogenesis	<i>Cre13.g573050</i>
Trough day		Monocarboxylic acid metabolic process	<i>AT4G25700</i>		Ribosome biogenesis	<i>ostta06g01560</i>		RNA metabolic process	<i>Cre01.g012350</i>
Peak dawn	811	Response to auxin	<i>AT7G29440</i>	1,744	Tetrapyrrole metabolic process	<i>ostta17g00130</i>	2,869	Tetrapyrrole metabolic process	<i>Cre06.g306300</i>
Trough dusk		Carbohydrate biosynthetic process	<i>AT7G70290</i>		Carbohydrate catabolic process	<i>ostta01g03040</i>		Carbohydrate catabolic process	<i>Cre01.g006950</i>
Peak dawn	132	Regulation of transcription	<i>AT5G62430</i>	739	Protein localization to membrane	<i>ostta03g05390</i>	1,212	Protein targeting to ER	<i>Cre16.g683950</i>
Trough night		Organic cyclic compound metabolic process	<i>AT7G20330</i>		Nucleotide metabolic process	<i>ostta15g00150</i>		Cyclic nucleotide metabolic process	<i>Cre02.g074150</i>
Peak day	980	DNA replication	<i>AT4G02060</i>	997	DNA metabolic process	<i>ostta03g03780</i>	1,987	Protein catabolic process	<i>Cre12.g531100</i>
Trough dawn		Actin cytoskeleton organization	<i>AT7G13180</i>		Establishment of localization	<i>ostta02g01890</i>		Golgi vesicle transport	<i>Cre10.g447350</i>
Peak day	127			141	Protein localization to membrane	<i>ostta01g00410</i>	256	Peptide biosynthetic process	<i>Cre10.g421600</i>
Trough day					Response to stress	<i>ostta08g03450</i>		Vitamin metabolic process	<i>Cre02.g090150</i>
Peak day	222	Tricarboxylic acid metabolic process	<i>AT5G04950</i>	514	Tetrapyrrole biosynthetic process	<i>ostta02g02380</i>	962	Photosynthesis	<i>Cre10.g425900</i>
Trough dusk		Isoprenoid metabolic process	<i>AT4G15560</i>		Photosynthesis	<i>ostta14g00150</i>		Tetrapyrrole biosynthetic process	<i>Cre06.g306300</i>
Peak day	935	Photosynthesis	<i>AT4G10340</i>	1,060	Signal transduction	<i>ostta04g01810</i>	1,467	Photosynthesis	<i>Cre16.g673650</i>
Trough night		Ribosome biogenesis	<i>AT5G16750</i>		DNA replication	<i>ostta01g02580</i>		Carbohydrate derivative	<i>Cre12.g508700</i>
Peak dusk	907	DNA metabolic process	<i>AT2G42120</i>	1,620	Photosynthesis	<i>ostta14g00150</i>	3,751	Biosynthetic process	<i>Cre01.g017450</i>
Trough dawn		Carbohydrate catabolic process	<i>AT2G32290</i>		DNA metabolic process	<i>ostta02g00690</i>		DNA metabolic process	<i>Cre12.g508700</i>
Peak dusk	432	Vitamin metabolic process	<i>AT2G29630</i>	437	Microtubule-based process	<i>ostta10g00680</i>	1,573	Carbohydrate biosynthetic process	<i>Cre02.g086650</i>
Trough day		Peroxisome organization	<i>AT1G47750</i>		Vesicle-mediated transport	<i>ostta09g03760</i>	882	Mitotic cell cycle	<i>Cre02.g086650</i>
Peak dusk	196	Peptide biosynthetic process	<i>AT4G29430</i>	696	DNA replication	<i>ostta04g04640</i>		Microtubule-based process	<i>Cre10.g456350</i>
Trough night					RNA processing	<i>ostta11g03270</i>		Vesicle-mediated transport	<i>Cre02.g087551</i>
Peak night	1,014	Dephosphorylation	<i>AT1G71860</i>	1,234	Protein modification process	<i>ostta11g02830</i>	1,488	Chromatin organization	<i>Cre16.g668200</i>
Trough dawn					Ribosome biogenesis	<i>ostta09g00820</i>		Phosphorylation	<i>Cre10.g423200</i>
Peak night								Nucleotide biosynthetic process	<i>Cre13.g606250</i>
								Protein metabolic process	<i>Cre01.g035500</i>

(Continued)

TABLE 3 | Continued

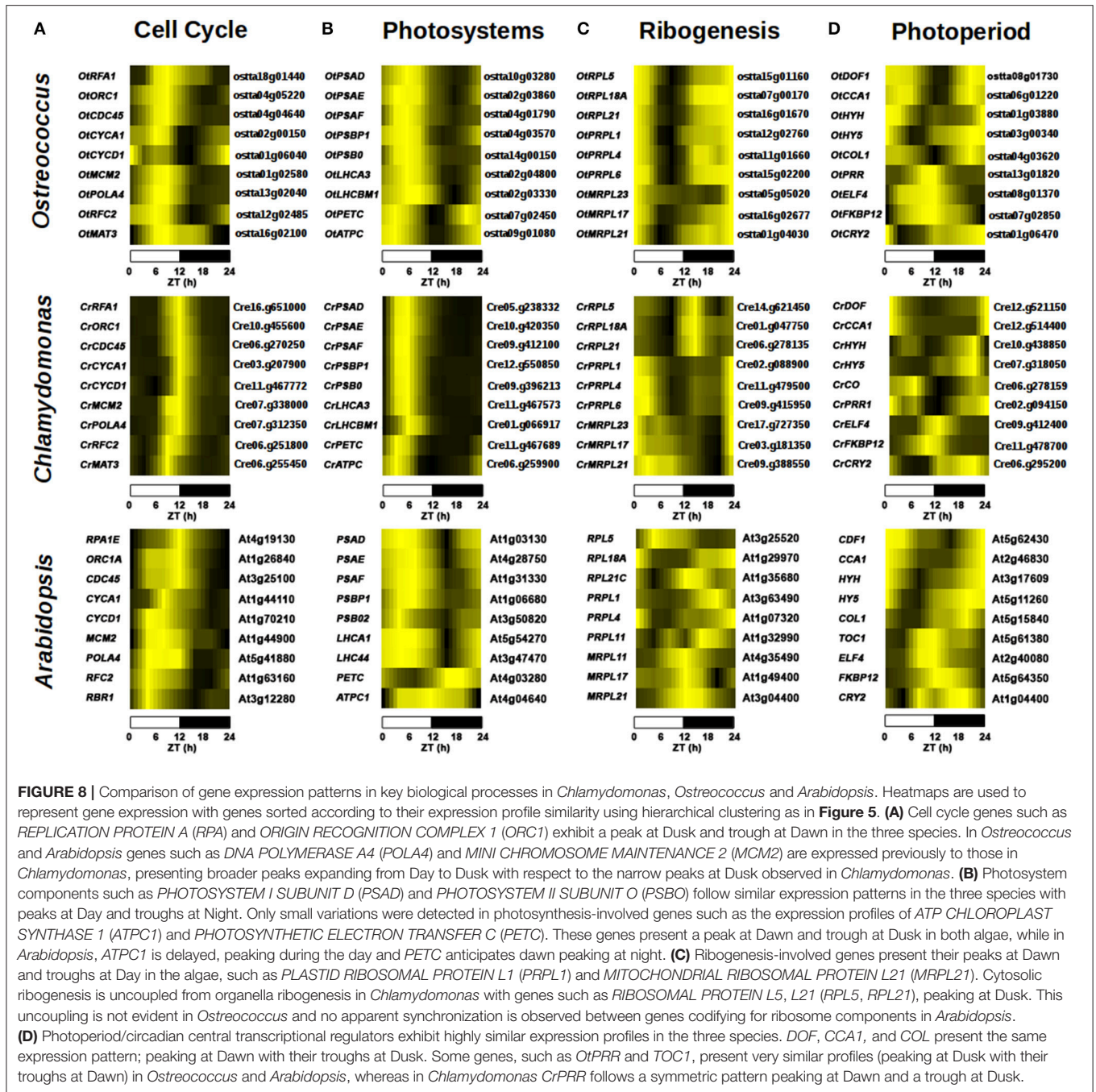
	<i>Arabidopsis thaliana</i>			<i>Ostreococcus tauri</i>			<i>Chlamydomonas reinhardtii</i>		
	Size	Enriched GO terms	Genes	Size	Enriched GO terms	Genes	Size	Enriched GO terms	Genes
Trough day		Autophagy	AT3G60640		Vitamin metabolic process	ostta09g00820		Regulation of cellular process	Cre12.g489000
Peak night	177	Hexose metabolic process	AT5G35790	689	Gene expression	ostta18g01220	951	Protein metabolic process	Cre01.g001400
Trough dusk		Response to stimulus	AT1G10470		Peptide biosynthetic process	ostta02g01970		Signal transduction	Cre13.g571200
Peak night	3			60	L-serine metabolic process	ostta12g03060	57	Cellular respiration	Cre13.g568800
Trough night									

only significantly detected in the gene cluster with peak at Day and trough at Night (Table 3). A closer inspection into ribogenesis in *Chlamydomonas* reveals an uncoupling between the genesis of cytosolic, plastid and mitochondrial ribosomes (Zones et al., 2015). Cytosolic ribosome components peak at Dusk (i.e., *CrRPL5*, Cre14.g621450), plastid ribosome components peak at Dawn (i.e., *CrPRPL1*, Cre02.g088900) and mitochondrial ribosome component present a broad peak at Dawn (i.e., *CrMRPL21*, Cre09.g388550) (Figure 8C). This uncoupling is less evident in *Ostreococcus* where genes codifying for components of cytosolic (i.e., *OtRPL5*, ostta15g01160), plastid (i.e., *OtPRPL1*, ostta12g02760) and mitochondrial (i.e., *OtMRPL21*, ostta01g04030) ribosomes exhibit a peak at Dawn and a trough at Day (Figure 8C). No apparent synchronization is observed between genes codifying for ribosome components in *Arabidopsis* (Figure 8C). Therefore, the analysis shows how the light/dark cycles in ancient algae synchronized all ribosome synthesis at the same time point. In *Chlamydomonas* ribogenesis was divided into three different stages depending on the nuclear, chloroplast or mitochondrial genome control, while *Arabidopsis* showed a ribosome biogenesis almost independent from daily rhythms.

The conservation and evolution of the central transcriptional regulators in circadian/photoperiod response in *Arabidopsis* was analyzed. A strong conservation was observed between the expression profiles of the key circadian/photoperiod regulators in *Ostreococcus* and *Arabidopsis* with some variations in *Chlamydomonas* (Figure 8D). For instance, the three different species showed similar expression profiles of two central photoperiodic genes; *DOF* (Cre12.g51440, ostta04g02850, At5g62430) and *COL* (Cre06.g278159, ostta04g03620, At5g15840) peaking at Dawn, with their trough at Dusk (Figure 8D). Remarkably, two of the central genes in the circadian clock; *CCA1* (At2g46830, ostta06g01220) and *TOC1* (At5g61380, ostta13g01820) exhibit the same symmetric expression profiles in *Ostreococcus* and *Arabidopsis*, while in *Chlamydomonas* the putative orthologues *CrCCA1* (Cre12.g514400) and *CrTOC1* (Cre02.g094150) detected in our analysis present the same expression profile, peaking at Dawn with their troughs at Dusk (Figure 8D). This suggests an independent and divergent evolution in *Chlamydomonas* of the core regulatory circuit of circadian rhythms as previously suggested (Mittag et al., 2005). Nevertheless, the full validation of these results would require more extensive analysis and experimental work.

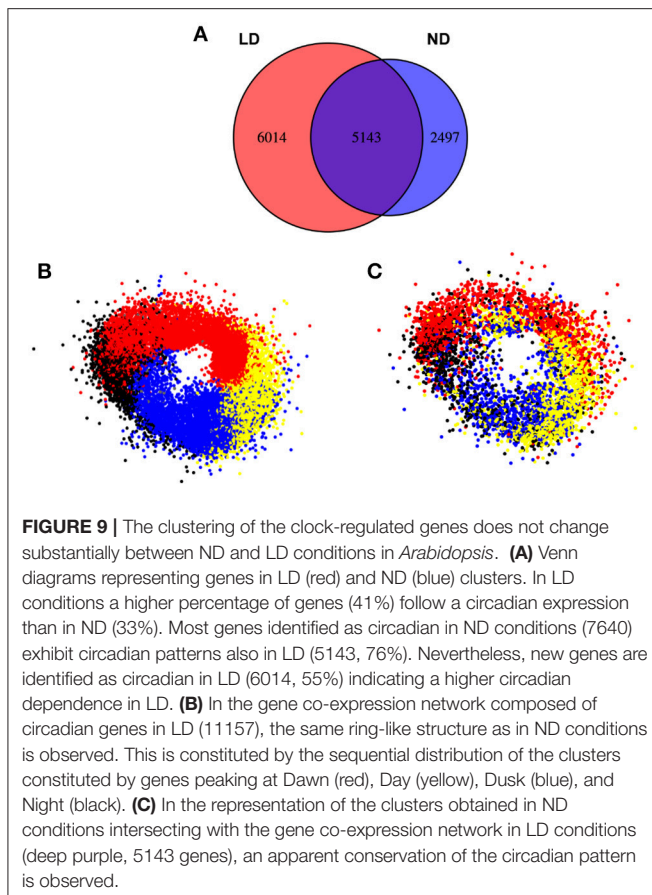
Other relevant genes involved in photomorphogenesis such as the bZIP TF *ELONGATED HYPOCOTYL 5* (*HY5*) and *HY5-HOMOLOG* (*HYH*) exhibit the same expression pattern in *Chlamydomonas* (Cre07.g318050, Cre10.g438850), *Ostreococcus* (ostta03g00340, ostta01g03880), and *Arabidopsis* (At5g11260, At3g17609), peaking at Dawn with their troughs at Day (Figures 8D, 11C,D).

It was also interesting to observe how the clustering approach worked when the photoperiod changed from ND to LD (Figure 9). This way, when the expression profile by RNAseq of a 24 h Col-0 plant grown in LD (red) was compared with the profile in ND (blue), more than 41% of the total genes



expressed in LD presented a daily rhythmic profile (11157), while the percentage was reduced to 37% of genes in ND (7640). Out of the 7640 genes showing a periodic regulation in ND, 5143 (deep purple, 67%) could be also identified in LD, while approximately half of the 11157 genes in LD (6014, 54%) did not follow a daily rhythmic pattern in ND. This could mean that increasing the photoperiod also augments the daily rhythmic regulation of the transcriptome in higher plants, although never reaching the numbers shown in microalgae. When we organized the periodic-expressed genes in clusters, based on their peak

and trough during the day as in **Figure 8**, a clear clock-wise distribution was observed (**Figure 9B**). This seems to indicate that the basic organization of the daily rhythms is independent of the photoperiod. However, when a temporal-clustered gene co-expression network was constructed with the intersection genes between daily rhythmic genes in LD and ND (5143 genes, deep purple) a diffused circadian pattern of expression was observed, suggesting a displacement or alteration in the daily rhythmic expression of a substantial number of genes due to the extended photoperiod (**Figure 9C**).



Module and Pathway Conservation Reveals High Level of Conservation between *Ostreococcus* and *Chlamydomonas* and a Moderate Level of Conservation between Microalgae and *Arabidopsis*

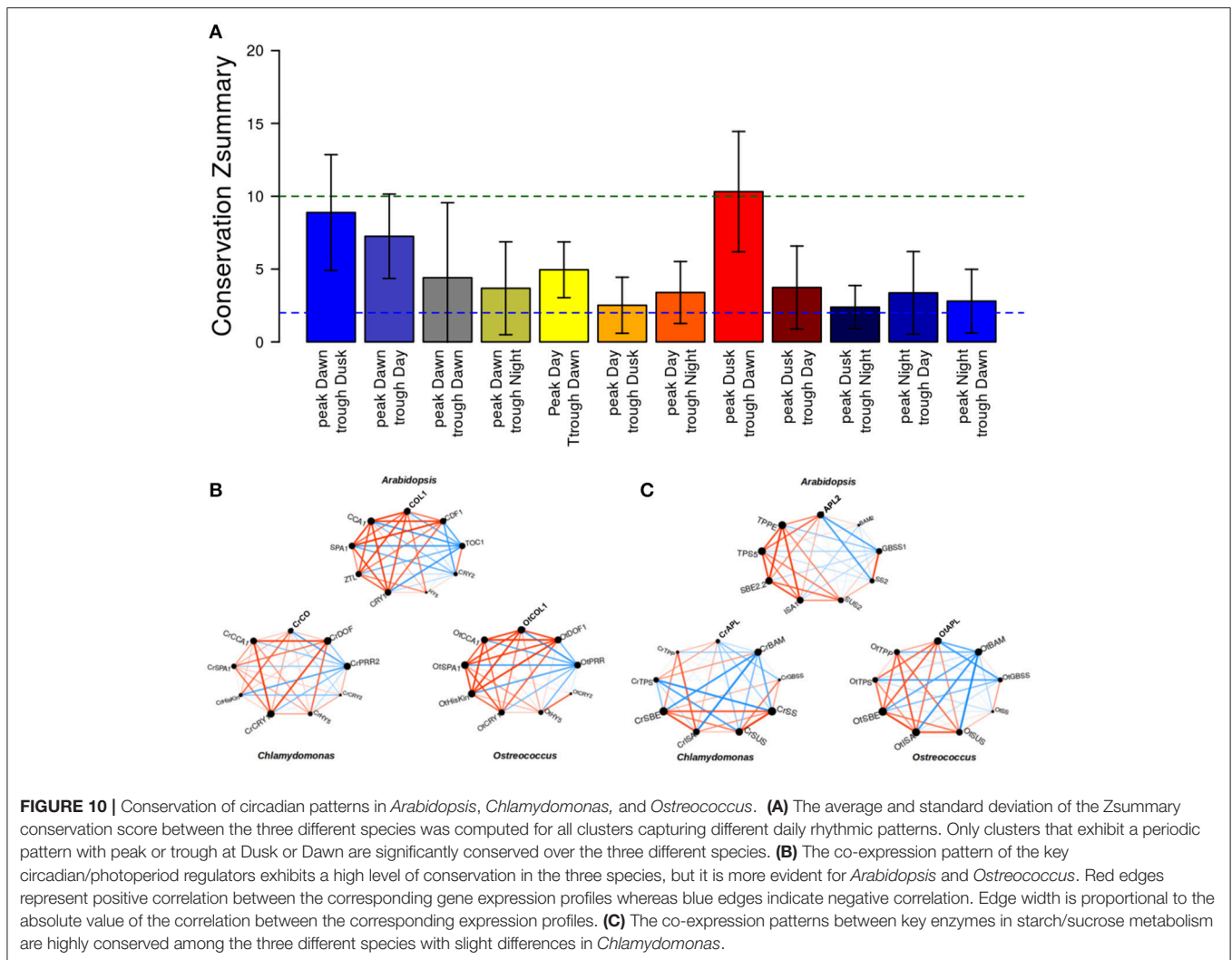
With the aim of determining highly likely functional orthologues between the three different species, information regarding co-expression patterns was integrated with the results obtained previously using the MBBH method. Similar approaches based on the integration of sequence similarity, gene expression profiles and co-expression patterns have been recently successfully applied to determine functional orthologues (Romero-Campero et al., 2013; Das et al., 2016). Thus, it was assumed that two MBBH potential orthologous periodic genes from two different species exhibited a conserved daily pattern when both presented their peak and trough in the same time interval (i.e., same circadian cluster) or when the Pearson correlation coefficient between their expression profiles was higher than 0.98. Therefore, such two genes could be named as expresologues (Das et al., 2016).

According to this criterion approximately 34% of *Arabidopsis* daily rhythmic genes presented an expresologue in *Ostreococcus* or *Chlamydomonas*. This suggests a high level of conservation in spite of the large evolutionary distance

between flowering plants and microalgae and presumably, along the plant evolutionary lineage. Interestingly, only the *Arabidopsis* merged cluster of daily rhythmic genes peaking at Dusk was significantly enriched in *Ostreococcus* and *Chlamydomonas* expresologues (p -value of 4.84×10^{-5} by Fisher's exact test), suggesting that most of these conserved genes maintain a strong influence of light/dark transitions on their expression control. In *Ostreococcus*, 83.62% of the daily rhythmic genes present an expresologue in *Chlamydomonas* but just 36.16% have an expresologue in *Arabidopsis*. In *Chlamydomonas*, 52.71% of the daily rhythmic genes present an expresologue in *Ostreococcus* whereas only 19.05% have an *Arabidopsis* expresologue, which supports a more divergent evolution in this species when compared to *Ostreococcus*.

To study the conservation of daily rhythmic patterns in the different clusters among the three species, beyond the comparison between individual gene expression profiles, the Summary Composite Conservation Statistic (Zsummary) was computed as defined in Langfelder et al. (2011) (Figure 10A). A Zsummary value lower than 2 indicates no conservation, a Zsummary value 2–10 implies a moderate conservation, while Zsummary greater than 10 constitutes evidence of a great level of conservation. For each daily rhythmic gene cluster, the Zsummary for the six different possible comparisons (*Arabidopsis* vs. *Ostreococcus*, *Arabidopsis* vs. *Chlamydomonas*, *Ostreococcus* vs. *Arabidopsis*, *Ostreococcus* vs. *Chlamydomonas*, *Chlamydomonas* vs. *Arabidopsis* and *Chlamydomonas* vs. *Ostreococcus*) was computed and the corresponding average and standard deviation was plotted. In Figure 10A, it can be observed that clusters with a peak at Dawn present a moderate level of daily rhythmic pattern co-expression conservation. In fact, the highest level of conservation was obtained for the cluster with peak at Dawn and trough at Dusk and the cluster peak at Dusk and trough at Dawn (Zsummary > 10). This suggests that evolution has mostly conserved periodic genes with a strong influence from the light/dark and dark/light transitions in their expression profiles, again hinting the importance of these transitory states in plant physiology and their conservation across the entire green lineage.

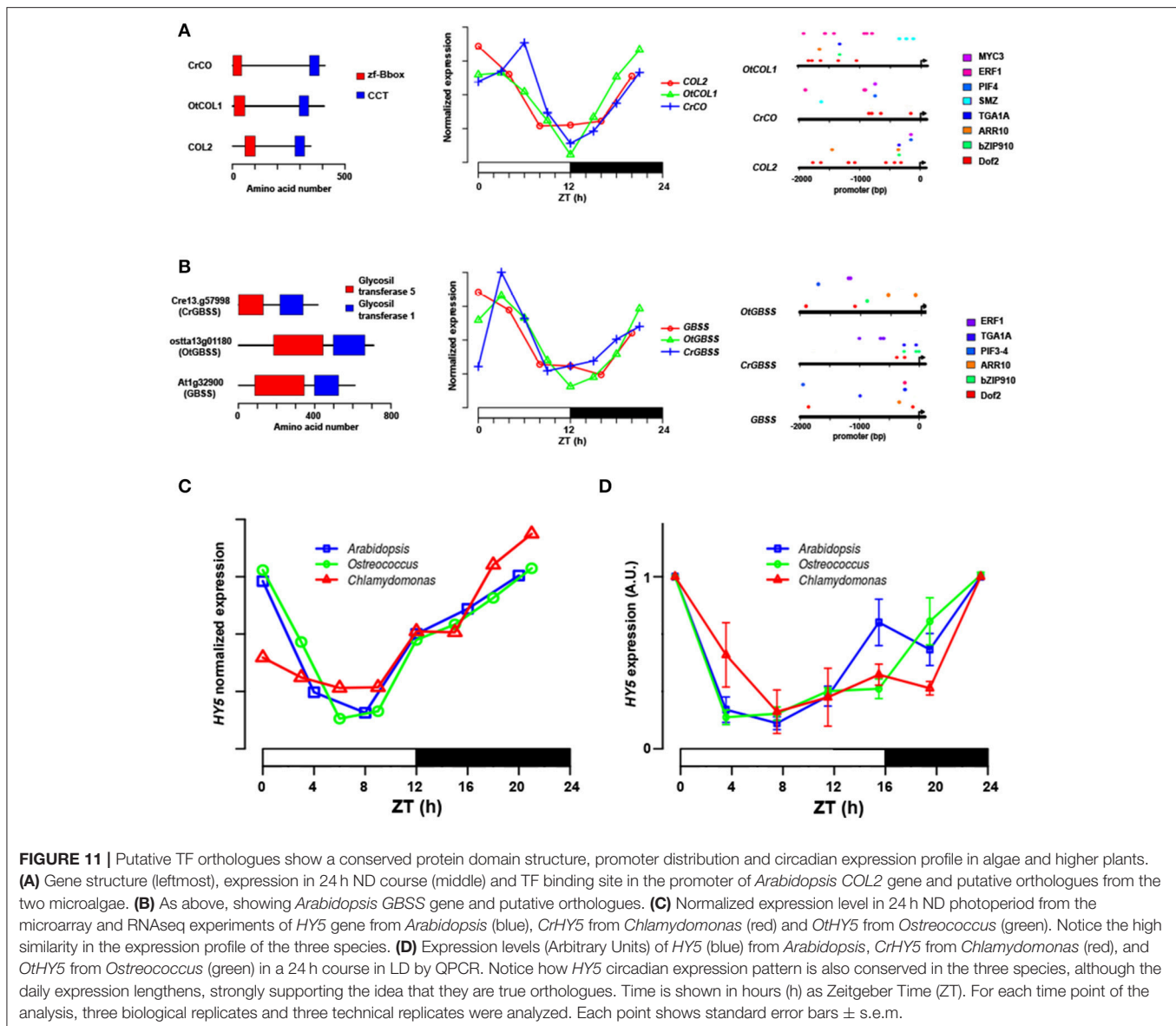
To corroborate the conservation of co-expression patterns in the three different species, two central pathways identified as daily-regulated in *Arabidopsis* were chosen (Table 2). For these gene sub-clusters, smaller gene co-expression networks were constructed, namely key regulators in circadian/photoperiod system (Figure 10B) and central enzymes in starch/sucrose metabolism (Figure 10C). In these networks, where red edges indicate positive correlations and blue edges indicate negative correlations, a general conservation between the plots can be observed (Figures 10B,C). The circadian/photoperiod plot revealed a high conservation among the three sub-clusters. Nevertheless, a general higher conservation between *Arabidopsis* and *Ostreococcus* display, when compared to *Chlamydomonas* could be observed, indicating again that this microalgae has slightly differentiated from the core *Arabidopsis* model clock (Figure 10B). On the contrary, daily-regulated starch metabolic



genes showed a higher conservation degree between *Ostreococcus* and *Chlamydomonas*, maybe revealing that through the course of evolution, additional regulatory steps were incorporated into the higher plant starch synthesis control (Figure 10C). On the whole, this sub-cluster visualization allowed us to compare different subset of genes that had showed a correlation in our general analysis (Supplementary Tables 1–3) further indicating levels of conservation and divergence among the species and constituting efficient tools to initiate the study on the evolution of particular processes.

To further test the veracity of our approach, the promoters of genes showing a high level of conservation both in sequence similarity (Figures 11A,B, leftmost) and co-expression patterns (Figures 11A,B, middle) in the three species were analyzed. As observed in Figures 11A,B rightmost, the promoters of *COLs* (Figure 11A) and *GBSSs* (Figure 11B) orthologues showed also conservation in the TF binding sites identified in the corresponding gene promoters. This result suggests that through evolution, in order to maintain the regulation of these functional networks, whole set of genes have to

evolve synchronously and reveal why these networks are so resilient to change even in long evolutionary time scales. Finally, in order to provide an independent validation of the conservation of these daily rhythmic patterns and the capacity to predict orthology in modern plants and algae, the normalized expression of the *Arabidopsis* bZIP gene *HY5* and the putative orthologues detected in this study in *Chlamydomonas* (*CrHY5*) and *Ostreococcus* (*OtHY5*) from the ND microarray analysis (Figure 11C) were plotted and compared with the expression detected in a 24 h LD course by QPCR experiments (Figure 11D). Both plots showed a remarkable similarity, except that the LD expression profiles of *HY5* and orthologues prolonged the daily minimal expression levels compared to the ND profile until the last hours of light, when the mRNA levels started to rise again. Therefore, the expression profiles of *HY5* and orthologues showed a very clear conserved pattern, as had been shown in our previous analysis (Figure 8) indicating that the three genes have a high chance of expressing functional orthologues, a line of research that is now being followed in the laboratory.



CONCLUSIONS

As the amount of transcriptomics and phylogenomics data accumulates, the complexity of the evolution of the mechanisms governing the control of gene expression in eukaryotes becomes more evident. A paradigm of a specific pattern that controls the transcriptome occurs under alternating light/dark cycles (Millar, 2016). In photosynthetic organisms this control is particularly important due to their dependence on sunlight for most of their main physiological processes and thus, a complex periodic gene expression mechanism can be found as early as in microalgae (Mittag et al., 2005; Corellou et al., 2009). Here, a Systems Biology approach has been used to dissect this dependence and find out mechanisms anciently controlled by daily rhythms and those that have acquired a new regulation. To achieve this, MBBH tool was developed to help discern orthology from species

evolutionarily very distant as algae and modern plants. When used in parallel with gene co-expression networks, a very strong toolset to understand how different processes have conserved a periodic regulation through long evolutionary distances has been built. Furthermore, a web-based tool has been implemented to allow the study of the daily rhythmic evolution of any set of genes from algae to modern plants in a similar way to previous web-based tools that allow researchers to compare diurnal expression profiling between monocots and dicots (Mockler et al., 2007). Together with the available pipelines and databases presented in former papers (Romero-Campero et al., 2013, 2016) they constitute strong resources for the research community.

Using these tools, we confirmed that primitive picoeukaryote microalgae, such as *Ostreococcus*, govern most of its transcriptome (a practical 100%) by a daily rhythmic mechanism under alternating light/dark cycles (Monnier et al., 2010). Other

microalgae have reduced this light dependence, but the number of genes regulated by light is still significantly high. This is the case of *Chlamydomonas* (Harris, 2001), which can perform some complex physiological processes independently of light/dark cycles, now controlled by other factors such as external biotic stimuli. Following a similar tendency, *Arabidopsis* has reduced the temporal dependence of its gene expression control under light/dark cycles to just over a third of its transcriptome (Michael et al., 2008) but is still significantly higher than the 10–15% of genes showing a periodic control in mammals (Lowrey and Takahashi, 2011). Nevertheless, it has been reported that up to 89% of the *Arabidopsis* transcriptome could follow a daily rhythmic pattern when different environmental signals, such as temperature, are combined with light/dark cycles (Michael et al., 2008). Therefore, plants seem to possess several molecular mechanisms to integrate diverse environmental signals beside light/dark cycles to exert daily rhythmic regulation over their transcriptome that are missing in microalgae. Temperature seems to be a key factor among these signals. The lack of potential microalgal orthologues for key genes in the evening loop such as GI and ZTL, that exert the temperature compensation to the clock in *Arabidopsis*, could explain these differences.

Finally, the clustering analysis performed attending to gene rhythmic expression has shown that the co-expression networks in the three organisms are distributed in the same order as the clock, giving the circadian timer a real associative and temporal dimension that participates in the coordination of different physiological processes such as ribosome synthesis or starch metabolism. In algae, those clusters that showed maximum expression during the light/dark transitions are the most abundant, while in *Arabidopsis*, these clusters containing orthologous genes, seem to be expressed during the dark or light periods. It seems then that *Arabidopsis* has been able to predict dawn and dusk in a better way than algae and has advanced gene expression in order to anticipate these transitions. Moreover, when a global conservation analysis is performed, only genes with their peaks and troughs at dawn and dusk are strongly conserved between microalgae and plants, again suggesting a diversification in the temporal control of gene

expression during their evolution. Finally, we have also shown the powerful predictable capacity of the approach by picking up putative orthologues that may result in good candidates for further studies in the future. This approach could help to better understand daily rhythmic regulation in algal systems, which may be important to understand the more complex mechanisms of higher plants and shed some light on how these mechanisms have evolved in the green lineage.

AUTHOR CONTRIBUTIONS

Pd and FR performed bioinformatics analysis. Pd and MR carried out experimental work. FV, JR, and FR conceived the study, interpreted the results and wrote the paper. All authors read and approved the manuscript.

FUNDING

The authors would like to thank funding from projects BIO2011-28847-CO2-00 and BIO2014-52425-P (Spanish Ministry of Economy and Competitiveness, MINECO) partially supported by FEDER funding.

ACKNOWLEDGMENTS

The authors would like to acknowledge Rafael Álvarez-Romo for the maintenance and assistance with the use of the high performance computing facilities of the CIC-Cartuja Data Processing Center that have been instrumental to the development of this study. We acknowledge support of the publication fee by the CSIC Open Access Publication Support Initiative through its Unit of Information Resources for Research (URICI).

SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: <http://journal.frontiersin.org/article/10.3389/fpls.2017.01217/full#supplementary-material>

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Conflict of Interest Statement: 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.

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