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Citation for published version:
Edwards, KD, Akman, OE, Knox, K, Lumsden, PJ, Thomson, AW, Brown, PE, Pokhilko, A, Kozma-Bognar, L, Nagy, F, Rand, DA & Millar, AJ 2010, 'Quantitative analysis of regulatory flexibility under changing environmental conditions' Molecular Systems Biology, vol 6, 424. DOI: 10.1038/msb.2010.81

Digital Object Identifier (DOI):
10.1038/msb.2010.81

Link:
Link to publication record in Edinburgh Research Explorer

Document Version:
Publisher's PDF, also known as Version of record

Published In:
Molecular Systems Biology

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Quantitative analysis of regulatory flexibility under changing environmental conditions

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The circadian clock controls 24-h rhythms in many biological processes, allowing appropriate timing of biological rhythms relative to dawn and dusk. Known clock circuits include multiple, interlocked feedback loops. Theory suggested that multiple loops contribute the flexibility for molecular rhythms to track multiple phases of the external cycle. Clear dawn- and dusk-tracking rhythms illustrate the flexibility of timing in Ipomoea nil. Molecular clock components in Arabidopsis thaliana showed complex, photoperiod-dependent regulation, which was analysed by comparison with three contrasting models. A simple, quantitative measure, Dusk Sensitivity, was introduced to compare the behaviour of clock models with varying loop complexity. Evening-expressed clock genes showed photoperiod-dependent dusk sensitivity, as predicted by the three-loop model, whereas the one- and two-loop models tracked dawn and dusk, respectively. Output genes for starch degradation achieved dusk-tracking expression through light regulation, rather than a dusk-tracking rhythm. Model analysis predicted which biochemical processes could be manipulated to extend dusk tracking. Our results reveal how an operating principle of biological regulators applies specifically to the plant circadian clock.

Introduction

Most eukaryotes and some prokaryotes possess circadian clocks, which regulate ~24 h rhythms in metabolism, physiology and behaviour, allowing organisms to anticipate predictable changes in the day/night cycle (Bell-Pedersen et al., 2005). All known circadian clock mechanisms comprise surprisingly complex circuits of nested or interlocked feedback loops (Bell-Pedersen et al., 2005; Kitayama et al., 2008). Microarray studies in organisms from mammals (Panda et al., 2002; Ueda et al., 2002) to plants (Edwards et al., 2006; Covington et al., 2008; Michael et al., 2008) have shown that large numbers of genes are rhythmically expressed, and that functionally related genes are often co-regulated at specific times of the day. Light and temperature signals entrain the clock mechanism to set the circadian phase, which describes the timing of endogenous rhythms relative to the environmental cycle (Bell-Pedersen et al., 2005). Normal circadian timing benefits growth and survival (Ouyang et al., 1998; Dodd et al., 2005), most probably due to the regulation of biological processes to an optimum phase in the daily cycle. Coordinating biochemical activity with the timing of dusk and dawn could provide a particular benefit in the case of carbon metabolism (Dodd et al., 2005). Some of the carbon fixed by photosynthesis is stored in the chloroplasts as transitory starch, which is broken down to provide a source of sugars throughout the night, preventing starvation-induced inhibition of plant growth (Zeeman et al., 2007; Graf et al., 2010). Several genes involved in starch metabolism are rhythmically regulated (Harmer et al., 2000; Smith et al., 2004; Blasing et al., 2005; Edwards et al., 2006; Michael et al., 2008). Their expression profiles over a light/dark cycle
combine circadian control, direct regulation by light and indirect light regulation by sugar signalling and by circadian entrainment (Blasing et al., 2005; Usadel et al., 2008). Systems biology aims to support quantitative analysis, understanding and intervention in such complex, dynamic systems.

In temperate regions, the length of the day (photoperiod) changes markedly with the seasons. Many organisms use a photoperiod signal to time annual transitions in development, such as flowering or bud dormancy in plants and reproductive development in mammals and birds (Dunlap et al., 2004). The circadian clock underlies the measurement of day length for these important annual events (Bohlenius et al., 2006; Imaizumi and Kay, 2006; Hazlerigg and Loudon, 2008). Our focus here is on the timing of biological processes within the day/night cycle, rather than on the amount of a photoperiod-dependent response. The timing of circadian rhythms might be expected to respond to a changing photoperiod, in order to anticipate a particular phase of the day/night cycle robustly under many conditions. Consistent with this notion, the phase of particular circadian rhythms in plants has been shown to alter with the photoperiod (Millar and Kay, 1996; Love et al., 2004; Perales and Mas, 2007).

If multiple phases of the day/night cycle have adaptive significance, then an important question is how biological rhythms gain the flexibility to track each of the external phases, most obviously dawn and dusk, as their relative timing changes with the seasons. We previously used mathematical analysis to understand the design principles that might underlie the complex, interlocking feedback loop circuits that have been identified in all circadian clock mechanisms. Tunability of period under constant conditions has been proposed as one benefit from mixed feedback circuits (Tsai et al., 2008). Environmental noise was shown to favour loop complexity in clock systems evolved in silico to anticipate environmental transitions (Troein et al., 2009). Our analytical results showed that the presence of multiple negative feedback loops could increase the flexibility of a clock gene network, for example, permitting distinct regulation of multiple phases in light/dark cycles (Rand et al., 2004, 2006). We have recently shown that such flexibility can support increased robustness, as defined by Kitano (2007), if the flexibility is appropriately linked to environmental changes (Akman et al., 2010).

The clock gene network of the model plant Arabidopsis thaliana is based on a feedback loop involving two closely related transcription factors, CIRCADIAN CLOCK ASSOCIATED 1 (CCA1) and LATE ELONGATED HYPOCOTYL (LHY), and the pseudo-response regulator TIMING OF CAB EXPRESSION 1 (TOC1) (McClung, 2006). Understanding of the plant clock has been formalised progressively in mathematical models (Supplementary Figure 1; Locke et al., 2005a, b, 2006; Zeilinger et al., 2006). An initial model consisted of a single loop, in which a combined CCA1 and LHY protein repressed the expression of TOC1, which in turn induced the expression of CCA1/LHY (Locke et al., 2005a). Inability of this model to explain the experimental data in clock mutants led to its extension to include an interlocked, evening feedback loop between TOC1 and a hypothetical gene Y, and a morning feedback loop between LHY/CCA1 and the combined TOC1 paralogues PSEUDO-RESPONSE REGULATOR 7 (PRR7) and PRR9 (Locke et al., 2005b, 2006; Zeilinger et al., 2006). The model is highly light responsive: light signals activate transcription of LHY/CCA1, PRR7/9 and G genes and degradation of TOC1 protein. Model predictions and experimental evidence led to the proposal of GIGANTEA (GI) as a candidate for part of the Y function in the evening loop (Locke et al., 2006), and recent data confirm that GI alone does not account for Y function (Martin-Tryon et al., 2007; Ito et al., 2009). Additional gene-regulatory loops (McWatters et al., 2000; Hazen et al., 2005; Pruneda-Paz et al., 2009) and cytosolic signalling mechanisms (Dodd et al., 2007) have yet to be included in the models, and these may contribute further complexity to the plant clock.

The existence of coupled feedback loops in the plant clock opens the possibility for increased flexibility in the relative phase of clock components (Locke et al., 2006). This is analogous to the coupled ‘evening’ (E) and ‘morning’ (M) oscillators that allow the activity rhythms of nocturnal rodents to track the predicted times of dusk and dawn, respectively (Pittendrigh and Daan, 1976). Distinct groups of neurones exhibit E and M properties in Drosophila and in the mouse, with strong, intercellular coupling to combine their properties in the intact animals (Jagota et al., 2000; Stoleru et al., 2004; Inagaki et al., 2007). Plant cell clocks, in contrast, are only weakly coupled by circadian signals within each organ (Thain et al., 2000; Fukuda et al., 2007), although light signals can indirectly couple clocks in distant organs (James et al., 2008). Our understanding of the plant clock mechanism emphasises intracellular regulation, because it is based upon data for genes that are broadly expressed within aerial plant tissues. Tissue- and organ-specific modifications of the plant clock mechanism may provide an additional level of complexity to spatially distinct rhythms (Thain et al., 2002; París et al., 2007).

In this study, we combined experimental and theoretical approaches to determine how much the potential flexibility of the three-loop circuit has been exploited in the evolution of the actual circadian system in Arabidopsis seedlings. Expression profiles for the Arabidopsis clock genes were measured across multiple photoperiods, with new controls for the LUCIFERASE (LUC) reporter gene imaging methods. ‘Dusk sensitivity’ is introduced as a simple measure for the pattern of entrainment of any circadian rhythm, and is applied to reveal the distinct regulation characteristic to each of the Arabidopsis clock models. The in vivo data validated the structure and detailed behaviour of the evening loop in the three-loop clock model, and quantified the behaviour of morning genes for future models. Finally, the dusk sensitivity measure was extended to predict how the entrainment of a three-loop clock could be manipulated to extend dusk tracking.

**Results**

**Contrasting entrainment patterns in a model species for classical plant physiology**

‘Short-day’ plants such as Ipomoea nil (Pharbitis nil) trigger flowering under shortening day lengths (with correspondingly lengthening nights). Such species have long been known for ‘dusk-tracking’ entrainment (Heide et al., 1988; Thomas and Vince-Prue, 1997). Plants of I. nil grown in constant light were not induced to flower (as in light:dark cycles with a long photoperiod), until they were transferred to a test interval of
constant darkness that mimicked a long night. The circadian rhythm that controls flowering was measured by the repression of flowering in response to a ‘night-break’ light pulse (Figure 1). The time of maximum repression (NBmax) was completely determined by the time of the transfer to darkness, as other authors have shown (Lumsden et al., 1995; Thomas and Vince-Prue, 1997 and references therein). However, the peak times of output rhythms that peak in the day, such as transpiration rate (shaded symbols) and maximum inhibition of flowering by a red light pulse (NBmax, open symbols), measured in darkness after different light intervals in *I. nil*. Shaded area of plot, darkness; open area, light.

**Figure 1** Dawn- and dusk-dominant rhythms show flexible timing in *Ipomoea nil*. Peak times are shown for rhythms of LHCβ expression (filled symbols), transpiration rate (shaded symbols) and maximum inhibition of flowering by a red light pulse (NBmax, open symbols), measured in darkness after different light intervals in *I. nil*. Shaded area of plot, darkness; open area, light.

### Entrainment patterns of clock gene RNAs in Arabidopsis

To test this, the timing of clock gene expression was measured under various photoperiods in Arabidopsis, using quantitative PCR (Q-PCR) assays or reporter gene imaging *in vivo*. Figure 2 shows the accumulation of RNA transcripts for three clock genes during photoperiods between 3 and 18 h, followed by constant light (LL) or darkness (DD). The RNA expression profiles were generally advanced to earlier times during the shorter photoperiod treatments, though the detailed photoperiod dependence of the expression profiles varied among the RNAs. The rising portion of the CCA1 RNA profile at Zeitgeber Time (ZT, where ZTOh—dawn) 16–24 h appears earlier in shorter photoperiods. The timing of the increase changes by only 5 h, comparing 6–18 h photoperiods. The effect appears to be more striking due to the higher peak level of expression in shorter photoperiods (Figure 2A). CCA1 levels peaked at ZT20–24h; 18 h photoperiods caused a delay of about 4 h compared with 3 h photoperiods. The TOC1 profile had a broader peak, which is discussed below. The tendency for increased peak expression under shorter photoperiods was shared to different extents by TOC1 and GI RNAs. Peak GI expression moved from ZT6h under 6-h photoperiods to ZT8h under 9- and 12-h photoperiods, and to 8–10 h under 18-h photoperiods (8–10 and 32–34 h in Figure 2C). The 3-h photoperiod caused an unexpected, biphasic profile in GI. GI RNA peaked in the light at ZT2h (2 and 26 h are replicate time points, Figure 2C) and again in darkness at 6–8 h or in light at 30–32 h (Figure 2C; see Supplementary information).

Phase plane plots of the first cycle of 6-, 12- and 18-h photoperiod data showed the dynamic relationships among the genes more clearly (Supplementary Figure 2), supporting the proposed causal interactions but also highlighting potential exceptions. The shoulder of TOC1 RNA abundance at ZT16h–20h in the 6-h photoperiod, for example, survived higher expression of its repressor CCA1 than under longer photoperiods (Supplementary Figure 2C).

RNA expression can be directly regulated by light signalling during the light:dark cycles, which complicates interpretation of the profiles. Circadian regulation is revealed under constant conditions, in LL or DD, where the effects of the entraining light:dark cycles on the clock can be assessed. The times of dusk in the entraining cycles varied by 15 h. In contrast, the peak times of CCA1 and GI RNA fell within a 2–3-h time range in the subsequent cycle in LL (44–68 h in Figure 2A and C). The peak times for each RNA spanned a 4-h time range in DD (24–48 h in Figure 2B, D, F, I and J, and Supplementary Figure 3). The small range of peak times relative to dawn indicates that the gene expression patterns showed only a limited response to the lights-off signal. Entrainment overall was dawn-dominant, more similar to the transpiration and LHCβ rhythms in *I. nil* than to the flowering rhythm (Figure 1). Simulations of the two-loop model (Locke et al., 2005b) illustrate a contrasting, dusk-dominant entrainment: just shortening the photoperiod from 9 to 3 h was sufficient to cause a 3 h change in the simulated RNA peak times under LL (arrowheads in Figure 2G and H) and longer photoperiods caused even larger changes.

The TOC1 RNA showed characteristically broad peaks. A maximum at ZT8h was observed under 6-h photoperiods and subsequent LL (8 and 32 h, Figure 2E), although profiles in 3- and 9-h photoperiods suggested broader or later peaks at ZT10–12h. A later feature at ZT16–20h created a shoulder on the falling phase. The 3- and 9-h photoperiods were investigated in separate experiments with triplicate samples at 1-h time resolution over the TOC1 peak. The TOC1 RNA maximum occurred at ZT10h in 9-h photoperiods and at ZT11h in 3-h photoperiods (Figure 2I and J). Thus, the peak time was not advanced when the time of dusk advanced in short photoperiods, as it was in the two-loop model (Figure 2H), but was instead slightly delayed, reminiscent of the flexibility expected for the three-loop clock model (Locke et al., 2006). A feature at ZT8h and the shoulder at ZT16h were also suggested in these high-resolution time series (as indicated in the figure).

### Analysis of entrainment patterns in Arabidopsis clock models

The mathematical models of the Arabidopsis clock have contrasting behaviour in light–dark cycles (Locke et al., 2006),
and the underlying mechanisms are well defined. We therefore sought to understand the experimental data by comparison with the models. Timing rather than expression level was our focus. Expression levels in the models are arbitrary, because the data available during model construction had not allowed us to constrain the simulated expression levels. Time series from numerical simulation were analysed to find the time at which the simulated RNA levels (shown in full in Supplementary Figure 4) of each clock component reached their peak level, when the model was stably entrained to light–dark cycles (Figure 3B, D and F). The peak times showed contrasting patterns of entrainment for the RNAs (see below).

To show the entrainment patterns for both RNA and protein components in a compact form, we used dynamical systems perturbation theory to develop a measure of dusk sensitivity. The measure reflects how closely the peak and trough times match a change in the time of dusk, and is applicable to any entrained oscillator (see Supplementary information). A dusk sensitivity of 1 indicates that the clock component will perfectly track the time of dusk (strongly dusk-dominant entrainment), whereas a component with 0 dusk sensitivity will perfectly track dawn (strongly dawn-dominant entrainment). The measure is intended to follow naturally from the plots of Figure 3B, D and F, where the line joining data points for a dusk-tracking component

Figure 2 Arabidopsis clock gene expression changes with photoperiod. Transcript abundance measured with 2 h time resolution by Q-PCR relative to an ACT2 standard, for clock genes CCA1 (A, B), G1 (C, D) and TOC1 (E, F, I, J) after entrainment to 24-h light:dark cycles (LD), including a photoperiod of 3 h (red), 6 h (orange), 9 h (yellow/black), 12 h (green) or 18 h (blue). Samples were taken during one diurnal cycle and after release into constant light (LL; A, C, E) or darkness (DD; B, D, F). Time-points 0–22 h are identical for LL and DD. Time stamps below (H, J) apply to all panels. Error bars represent the range of biological duplicates (A–F) or the SE of triplicates (I, J). Light conditions for three photoperiods are shown (open bar, light interval; shaded bar, darkness, with colours orange for 6 h photoperiod, green for 12 h, blue for 18 h). Simulations of LHY (G) and TOC1 (H) RNA levels in the interlocking-loop model illustrate the large phase changes predicted by a dusk-responsive model under this range of photoperiods (3 h, red; 6 h, orange; 9 h, broader yellow; 12 h, green; 15 h, blue). Arrowheads in (G, H) highlight the 3-h phase shift between 3- and 9-h photoperiods. Time series for TOC1 expression in the 3-h (I) and 9-h (J) photoperiod followed by DD are shown with 1 h time resolution at the peaks, together with equivalent data replotted from (F). Arrowheads in (I, J) mark the complex peak waveform observed in the samples with higher time resolution. Source data is available for this figure at www.nature.com/msb.
has a gradient of 1, and for a dawn-tracking component has a gradient of 0.

Dusk sensitivity was computed for both peak and trough phases of each clock component for clock models with varying complexity under 12-h photoperiods (Figure 3A, C and E), and for comparison under 6- and 18-h photoperiods (Supplementary Figure 5). Where a component had multiple peaks or troughs per cycle, their dusk sensitivity was computed separately. The one-loop model (Locke et al, 2005a) has light input only to the morning component LHY/CCA1

Figure 3 Predicted and experimentally measured entrainment patterns in the clock of Arabidopsis. The one-loop (A, B), two-loop (C, D) and three-loop (E, F) models of the Arabidopsis clock were analysed to calculate the dusk sensitivity measure (A, C, E) for the peak (upward triangle) and trough (downward triangle) times of mRNA (m) and bulk protein (P) variables of all the genes in the model under a 12-h photoperiod. Dusk sensitivity close to 1 indicates dusk-dominant entrainment; close to 0, dawn-dominant entrainment. Where an expression profile has multiple peaks or troughs, its dusk sensitivity is plotted from left to right, in chronological order after dawn, with the convention that in each peak/trough pair the trough follows the peak in time. The models were solved numerically under a range of simulated photoperiods, resulting in the simulated RNA profiles plotted in Supplementary Figure 3. Times of the peak abundance for each simulated RNA during light:dark cycles are shown (B, D, F; see inset key for gene identity). For comparison, the peak expression time for six clock genes was measured in individual seedlings using mFourfit analysis of in vivo imaging data (Supplementary Figures 5 and 7) from transgenic plants carrying LUC reporter fusions under the same range of photoperiods (G), or following transfer to LL (H) or DD (I; see the inset key for gene identity). PRR9 is absent from I, because very low expression levels in DD prevented phase estimation. Shaded chart areas represent darkness. Error bars indicate the standard error of the mean, calculated as described in the Supplementary information. Two-way ANOVA on LD, LL and DD peak times showed a highly significant interaction between gene and photoperiod in each case (P < 0.001), indicating that the genes responded to photoperiod in significantly different patterns. Source data is available for this figure at www.nature.com/msb.
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( Supplementary Figure 1), and, in accord with theoretical predictions, entrainment of the whole model is locked to dawn (Figure 3A and B). In the two-loop model (Locke et al., 2005b), additional light input via Y in the evening loop allowed dusk-dominant entrainment for all components (Figure 3C and D). Only the acute, light-induced peaks of LHY/CCA1 and Y expression after dawn were dawn-dominant (Supplementary Figure 4B and G, and Supplementary information). Neither model changed its behaviour across the range of photoperiod (Supplementary Figure 5A, C and E, and data not shown).

The three-loop model includes an additional feedback loop with light input to PRR7/9 (Supplementary Figure 1C), which conferred strong dawn dominance to LHY/CCA1 and PRR7/9 components (Figure 3E and F). However, the evening loop components Y and TOC1 showed intermediate dusk sensitivity values, indicating more flexible regulation that responded to both signals in the 12-h photoperiod (Figure 3E). This behaviour was altered substantially under the 6- and 18-h photoperiods (Supplementary Figures 5B, D and F), leading us to investigate its mechanisms.

Expression of Y showed the light-induced peak at dawn in the three-loop model, followed by a circadian peak at ZT10h (Supplementary Figure 4H), which showed intermediate dusk sensitivity in the 12-h photoperiod (Figure 3E). Under 6- or 9-h photoperiods, lights-off occurred during this peak of Y expression (marked in Supplementary Figure 4H). The abrupt end of light-activated transcription curtailed the peak earlier under 6-h than under 9-h photoperiods (Figure 3F), leading to high dusk sensitivity (Supplementary Figure 5B). Under 15- and 18-h photoperiods, Y expression was ended by negative feedback from TOC1 in the model’s evening loop. This occurred fully within the light interval, so the timing of this peak was unaffected by lights-off (Figure 3F) and its dusk sensitivity was very low (Supplementary Figure 5F).

Simulated TOC1 RNA lacks direct light regulation in the three-loop model, but its peak time showed a complex response, advancing under 3–9-h photoperiods, then delaying in 15-h photoperiods (Figure 3F). Dusk sensitivity of the simulated TOC1 peak was negative in short photoperiods, reflecting the fact that the peak occurred earlier in response to later dusk (Supplementary Figure 5B). Simulations of the three-loop model (data not shown) confirmed the intuition that extended light activation of Y transcription, which indirectly activates TOC1 (Supplementary Figure 1C), was responsible for the faster rise in TOC1 expression in photoperiods up to 9 h (Supplementary Figure 4F), leading to the earlier peak times (Figure 3F and G). Photoperiods greater than 9 h prolonged the light input to Y, allowing later TOC1 peak times, until negative feedback from TOC1 protein in the evening loop repressed Y (data not shown). Consistent with these dusk-independent events, dusk sensitivity of the TOC1 peak was low under long photoperiods (Supplementary Figure 5F).

The three-loop model supported flexible entrainment, as the peak times of the clock components changed relative to each other in a photoperiod-dependent manner. None of the components was dusk-sensitive in all photoperiods, in contrast to the two-loop model. Testing the subtle changes in regulation predicted by the model required peak time estimates with higher resolution than our RNA data.

Testing model predictions with in vivo reporter gene assays

Clock gene expression was therefore measured using in vivo LUCIFERASE (LUC) imaging, which avoided the biological variation introduced by sampling different plants at each time point for RNA assays. A modified analytical method, mFourfit, was developed to measure the times of peak expression in entrained rhythms with complex waveforms (see Supplementary information). The longitudinal LUC data allow direct measures of the time of peak expression in individual plants, and the mean and variation of timing in a population, which is the relevant behaviour. The destructive RNA assays, in contrast, reflect the peak time of the average expression level in the population samples but do not allow direct measures of timing. Peak expression times were measured under light-dark cycles (LD) with photoperiods from 3 to 18h, or following release into LL or DD (Figure 3G–I). The patterns of LUC expression (Supplementary Figures 6 and 8) closely followed the cognate mRNA profiles (see Supplementary information). The patterns of TOC1 timing in vivo showed strong similarity to the predictions of the three-loop model (Figure 3G–I). In particular, the TOC1 reporter showed the earliest peak in 9-h photoperiods (Figure 3G), as predicted by this model (Figure 3F). The peak time in 3-h photoperiods was slightly delayed, consistent with RNA data (Figure 2). GI reporter activity showed a dusk-dominant peak in photoperiods up to 9 h, as predicted by the dusk peak of Y in the three-loop model (Figure 3F). The GI reporter peak time was delayed in photoperiods greater than 9 h, again as predicted. The three-loop model closely reflected the observed regulation of these evening-expressed genes.

The major peaks of LHY and CCA1 expression followed closely after dawn (Figure 3G), as the LHY/CCA1 peak does in the three-loop model (Figure 3F). These LUC reporter profiles also showed an increase in expression before dawn (ZT19h–24h), consistent with the RNA data for CCA1 (Figure 2A). The observed timing of this initial rise was altered by 3 h in the 6-h photoperiod relative to the 12-h photoperiod conditions, in both RNA and LUC data. This indicated greater dusk sensitivity than predicted by the three-loop model, but less than the 5-h change predicted by the two-loop model (Figure 2G; see also Supplementary Figures 4B and C). The PRR9 reporter’s peak time was dawn-dominant (Figure 3G), but less so than the combined PRR9/7 gene in the three-loop model, which tracked the time of dawn (Figure 3F). Expression profiles after release into constant light (LL) or constant darkness (DD) more clearly separated the effects of dawn and dusk (Figure 3H and I). Peak times of LUC expression were more variable among individual plants than during light-dark cycles. The rise of CCA1 and LHY before dawn formed a distinct peak in short photoperiods under LD, for example, in both RNA data (Figure 2A) and LUC profiles (Supplementary Figures 6 and 8). The DD data showed this feature under longer photoperiods, without the intervening light response at dawn. The peak in DD retained dusk sensitivity of about 0.4 in photoperiods from 6 to 18 h (~5-h delay under a 12-h change in the time of dusk; Figure 3I). The strong resetting effect of dawn was revealed by comparison with the data under LL (Figure 3H), where the peak in CCA1
and LHY around 27 h reflected the phase set by the previous dawn at 0 h. The dusk sensitivity of this peak in LL was therefore close to 0.

Though the expression profiles were generally more dawn-dominant after release into LL and more dusk-dominant in DD, there was significant variation among genes. The profiles other than LHY and CCA1 showed the earliest peak times in LL after 6–9 h photoperiods (Figure 3H). Thus, their phase was not determined exclusively by dawn-dominant LHY and CCA1, but was instead reminiscent of the flexible regulation of TOC1 in LD. In DD, the entrainment pattern of all the genes was more similar to LHY and CCA1.

Dusk controls evening gene expression in Arabidopsis

None of the core clock genes tested showed the strong dusk dominance across all photoperiods in LD, which had been clear in the flowering rhythm of I. nil (Figure 1). We therefore measured RNA expression profiles under 6-, 12- and 18-h photoperiods (Figure 4 and Supplementary Figure 9), for clock-controlled genes involved in biological processes that are known to respond to the lights-off signal, namely the mobilisation of starch reserves (genes STARCH EXCESS 4 (SEX4) and DISEPROPORTIONATING ENZYME 2 (DPE2); Chia et al., 2004; Niittyla et al., 2006) and the photoperiodic control of flowering (genes FLAVIN BINDING KELCH REPEAT 1 (FKF1) and CONSTANS (CO); Imaizumi and Kay, 2006). CO, SEX4 and DPE2 showed broad expression profiles that hampered the estimation of peak expression time. The half-maximum points of the rising and falling phases were more reliable phase markers (Figure 4A), and have been used in several other circadian studies (Khalsa et al., 1992; Roenneberg et al., 2005). The CCA1 RNA rising-phase marker was advanced in 6-h compared with 12-h photoperiods, reflecting the dusk-responsive rise discussed above. The falling-phase marker was dawn-dominant, as expected after resetting at dawn. GI and FKF1 showed a small delay in rising (~2 h) and falling phases (~2–3.5 h) in the 18-h relative to the 6-h photoperiod conditions (Figure 4A), consistent with previous reports (Fowler et al., 1999; Imaizumi et al., 2003). The rising phase of CO, SEX4 and DPE2 was more dawn-dominant than their falling phase, which was strongly dusk-dominant (Figure 4B). Light signals were important for DPE2 and SEX4 profiles in LD, because both genes showed noisy peaks in LL and very low expression in DD that hampered detailed interpretation of timing. Both genes had previously been scored as rhythmic in LL (Edwards et al., 2006; Covington et al., 2008; Michael et al., 2008), but the clearly dusk-dominant fall in expression was only observed in DD, suggesting that this was a direct light response (Supplementary Figure 9). Timing of FKF1 and CO expression was more dawn dominant in LL than in DD (Supplementary Figure 9), suggesting that the time of lights-off had influenced the FKF1 and CO profiles under LD. The first rise of FKF1 and CO in DD also retained clear dusk responsiveness, suggesting that these genes, especially CO, were controlled by a circadian rhythm with dusk-responsive entrainment (Supplementary Figure 9).

Mathematical analysis of processes that control dusk sensitivity

The dusk sensitivity of circadian entrainment determines how clock-controlled genes will change their expression over the seasons. Designing a specific manipulation to the pattern of entrainment is non-trivial, because the pattern is an emergent property of the complex circadian system and all its light inputs. We therefore extended the dusk sensitivity measure to predict which biochemical processes could produce a desired change in the entrainment pattern of any component in the model, using the novel mathematical analysis described in the Supplementary information. As a test case, we sought manipulations that would alter the dusk sensitivity of the three-loop model under a 14-h photoperiod, because this is the condition when feedback in the evening loop limits the dusk sensitivity of the peak in TOC1 RNA (as discussed above). Processes in the evening loop, including light input to Y, were intuitively expected to affect the model’s dusk sensitivity, but none was as effective as the transcription of PRR7/9 (Supplementary Figure 11). This morning loop component represses LHY/CCA1, thereby indirectly activating both Y and TOC1 (Supplementary Figure 1C). Simulating a modification of PRR7/9 transcription in the three-loop model validated the analytical prediction: several clock components were converted from dawn to dusk dominance, including the time of
peak TOC1 RNA expression (Supplementary Figure 12). The extended analysis can thus prioritise the targets for future experimental investigation.

Discussion

By testing three mathematical models of the Arabidopsis clock, we showed that increased model complexity allowed the circadian phase of the model components to change flexibly in response to the photoperiod of the entraining LD cycle. The morning and evening feedback loops in the three-loop model, in particular, allowed for variation in the phase relationship between the different loops, enabling dawn- and dusk-dominant entrainment within a single gene network. Dusk dominance of the evening genes was predicted only for photoperiods between 6 and 12 h (Figure 3F), but this photoperiod range is physiologically relevant for Arabidopsis photoperiods between 6 and 12 h (Figure 3F), but this dominance of the evening genes was predicted only for dominant entrainment within a single gene network. Dusk dominance of the evening genes was predicted only for photoperiods between 6 and 12 h (Figure 3F), but this photoperiod range is physiologically relevant for Arabidopsis (Wilczek et al., 2009). This result is analogous to the observed dawn- and dusk-dominant entrainment of the morning and evening oscillators in neural clocks (Jagota et al., 2000; Stoleru et al., 2004; Inagaki et al., 2007). It illustrates the broader principle that multiple feedback loops can increase the flexibility of oscillator models (Rand et al., 2004, 2006). The same principle likely applies more generally to the complexity observed in non-oscillatory biological regulators: testing this hypothesis will require suitable mathematical models combined with experiments that manipulate relevant parameters, like the photoperiod variation tested here.

The three-loop model predicted much of the regulatory behaviour that we observed in gene expression data in vivo, although the model was constructed only using data for 12-h photoperiods and constant darkness (Locke et al., 2006). Dusk-dominant behaviour of the evening genes was observed over the predicted photoperiod range (Figures 2 and 3). The circuit’s flexibility was most evident in short photoperiods (<9 h; Figure 3F and G), where the peak of GI expression tracked dusk, the peaks of CCA1 and LHY expression were locked to dawn, and the TOC1 peak time moved slightly later as the time of dusk moved earlier. This response of TOC1 had not been identified before, as previous studies did not test very short photoperiods (Perales and Mas, 2007; Michael et al., 2008). The model’s correct and comprehensible prediction for the evening genes indicates the benefits of mathematical modelling in experimental design, and in understanding the interacting factors that control dynamic gene networks.

The model’s behaviour can be conceptually divided into the effects of light inputs, the day-time effects of the morning loop upon the evening-expressed genes, and the night-time effects of the evening loop upon the morning-expressed genes. The match to data indicated that the three-loop model recapitulated the day-time effects more accurately than the night-time effects. In particular, the GI and TOC1 data validated the location of light input to GI and the circuit structure of the model’s evening feedback loop. It does not necessarily follow that the GI and TOC1 proteins are the sole biochemical constituents of the equivalent functions in vivo (Locke et al., 2006; Martin-Tryon et al., 2007; Ito et al., 2009). We have previously argued that GI cannot constitute all of the Y function, for example (Locke et al., 2006). The light input at dawn dominated the behaviour of the morning loop components in the three-loop model, LHY/CCA1 and PRR9/7, whereas the data showed a significant dusk responsiveness in the regulation of LHY and CCA1 before dawn. This must reflect the functions of their regulator(s), represented by the activator X and/or the inhibitor PRR9/7 in the models. TCP21 (CHE) might contribute to this effect for CCA1 (Pruneda-Paz et al., 2009), but its mutant phenotype indicates that it alone is not X. The contribution of PRR9/7 to this effect will be re-evaluated elsewhere, in the light of emerging data.

The observed timing of the clock components was almost always more dawn dominant than dusk dominant, consistent with a strong resetting effect of dawn. Light induction of LHY/CCA1 at dawn is an important effect of light in the model, and also strongly affects the evening clock components. Light induction of LHY and CCA1 was detectable using the LUC reporters (Supplementary information and Supplementary Figure 6). Light induction of CCA1 and to a lesser extent LHY was directly demonstrated by RNA assays at high time resolution (Supplementary Figure 7), despite the light-induced destabilisation of the CCA1 RNA (Figure 2; Yakir et al., 2007). Our data allow direct comparison of the peak levels and rhythmic amplitudes of the clock genes on a broad scale (see Supplementary information), which have been little studied in any system. As many mechanisms affect absolute levels or amplitudes, we focussed on comparisons of timing, usually within a particular data type.

The clock genes provide markers for the core patterns of rhythmic regulation that are available to control thousands of circadian target genes. Under light–dark cycles, rhythmic transcripts detected by microarray assays showed peak times with two significant clusters, around dawn and around dusk (Michael et al., 2008). Comparing peak times on a gene-by-gene basis showed that many genes that were expressed up until dusk under 8- or 12-h photoperiods had later peak times under 16-h photoperiods or constant light (Figure 6B and D of Michael et al., 2008). This might appear paradoxical, given the absence of dusk-dominant profiles in the clock components. However, direct light signalling and indirect light effects via photosynthetic metabolism are important regulators during light:dark cycles (Blassing et al., 2005; Usadel et al., 2008). Transcriptome profiles under constant light showed a more even spread of peak times (Harmer et al., 2000), though two preferred phases were noted (Edwards et al., 2006; Covington et al., 2008; Michael et al., 2008). The evening-peaking pattern in the array data in LD is consistent with the regulation of evening genes such as SEX4 and DPE2 in our data, where each day’s light input sets the gene’s particular profile and especially its falling phase (Figure 4B and Supplementary Figure 9). Such evening-expressed genes will show dusk-tracking peak times in LD, but this cannot be taken as evidence for a dusk-tracking clock, even if the same genes show bona fide circadian regulation in LL. In 6–9-h photoperiods, for example, peak expression of GI tracks dusk in our data, as does X in the three-loop model. This reflects light regulation superimposed on a circadian rhythm (Fowler et al., 1999).

As the clock controls the plant’s responsiveness to light (Millar and Kay, 1996) and temperature (Fowler et al., 2005; Dodd et al., 2006) signals, the key function of the clock may be to balance direct and indirect environmental inputs to...
coordinate physiological and metabolic functions. Correct timing of rhythmic processes such as starch degradation is crucial for growth (Graf et al., 2010), so it is very likely that this temporal regulation of vegetative physiology has been subject to natural selection, and might contribute to crop improvement. Dusk sensitivity (Figure 3) and similar tools for targeted mathematical analysis of complex biological models help not only to understand the interacting genetic factors that contribute to temporal regulation, but also to prioritise the targets for manipulation (Supplementary Figures 11 and 12).

Natural genetic variation provides evidence (Bohlenius et al., 2006) of selection acting upon the dusk-dominant, photoperiod response rhythm in short-day plants (Figure 1). Strongly dusk-dominant rhythms seem less important for most gene expression profiles in Arabidopsis than the termination of light induction at dusk. However, the short-day species themselves show several dawn-dominant rhythms in gene expression and in processes involved in vegetative growth (Figure 1; Hayama et al., 2007). Conversely, the Arabidopsis photoperiod sensor gene CO showed an unusual degree of dusk-dominant, circadian control, apparently greater than its functional partner FKF1 (Figure 4B and Supplementary Figure 9). Additional mechanisms, such as further feedback loops, might of course affect the timing of specific genes in all cells. Alternatively, a cell-type-specific circadian clock that is modified for dusk dominance might exist in the phloem companion cells that express CO (An et al., 2004), without detectably affecting the results for other genes tested, because our assays averaged across all cells. Evidence for tissue-specific, and in particular vascular-specific, timing has been provided (Thain et al., 2002; Fukuda et al., 2007; Para et al., 2007; James et al., 2008). Differentiated regulation of the circadian clock might thus contribute to cell-type-specific processes, such as the photoperiod sensor, adding to the complexity of feedback loops within each cell.

Materials and methods

Plant materials and growth conditions

Seeds of I. nil (Pharbitis nil) Choisy cv. Violet were planted and grown as described (Lumsden et al., 1995). Unless otherwise noted, all Arabidopsis seeds were sterilised and grown on solid media containing 3% sucrose, as described previously (Edwards et al., 2005). Seedlings for Luciferase imaging were grown for 4 days at 22 °C in Sanyo MLR350 environmental test chambers (Sanyo, Osaka, Japan) under experimental photoperiods of 75 μmol m⁻² s⁻¹ cool white fluorescent light. Seedlings were then transferred to Percival I-380LL growth chambers (CLF Plant Climatics, Emersacker, Germany) at dawn on the 5th day and grown at 22 °C under an equal mix of Red and Blue LEDs at 20–30 μmol m⁻² s⁻¹ cool white fluorescent light. Seedlings were then released into LL or DD for 10 days. Details of the dusk sensitivity analyses are given in the Supplementary information. Each experiment included 22 individual seedlings of each genotype; experiments were replicated seven or more times per photoperiod for LD (four or more experiments released into LL conditions, and three or more into DD).

Measurement of photoperiodic response in I. nil

The method was essentially as described (Lumsden et al., 1995). Seedlings were grown in continuous light, and, after durations of between 72 and 96 h, ‘released’ into a dark period of 48 h. During the first 12 h of darkness, night breaks of 10 min red light were given at hourly intervals, to determine the time of maximal response to the night break. This time is termed NBmax.

Transpiration rate measurement

Plants of I. nil were grown as described above, and the mass of individual plants was recorded using a computerised system, to yield a time-series record of weight loss due to transpiration. The plant was weighed using Precisa 125A balances (Milton Keynes, UK) with automatic output via an RS232 port to an IBM PC-compatible computer. The system recorded the weight of each plant at 10-min intervals, from up to six balances in each experiment. Each recorded value is the mean of 10 individual measurements.

Measurement of gene expression

For RNA measurements in I. nil, plants were grown as described above. Total RNA was isolated as described (Kolar et al., 1995). The abundance of CAB (CO2) transcripts was determined on RNA gel blots using 20 μg of total RNA per sample, and hybridised to a radio-labelled tomato LHCl type1 CABI probe as described (Kolar et al., 1995). For quantitative RT–PCR in Arabidopsis, seedlings were harvested, and RNA was extracted and reverse transcribed as described previously (Locke et al., 2005b). Q-PCR was carried out using SYBR Green JumpStart Taq ReadyMix (Sigma, Gillingham, UK) in technical triplicate using the Relative Quantification function of a Light Cycler 480 (Roche, UK) to measure mRNA abundance. Expression values were normalised against ACTIN 2 (ACT2), TOC1, CCA1, LHY and GI primers have previously been described (Locke et al., 2005b; Edwards et al., 2006). See Supplementary information for other Q-PCR primer sequences.

Luciferase imaging was carried out as previously described (Gould et al., 2006) using Hamamatsu C4742-98 digital cameras operating at ~75°C under control of Wasabi software (Hamamatsu Photonics, Hamamatsu City, Japan). Imaging was started at dawn on the 6th day of growth. Bioluminescence levels were quantified using Metamorph software (MDS, Toronto, Canada) and phase estimates were produced with the mFourit function of BRASSv3 (Locke et al., 2005b; available at http://www.amillar.org; see Supplementary information). Each experiment included 22 individual seedlings of each genotype; experiments were replicated seven or more times per photoperiod for LD (four or more experiments released into LL conditions, and three or more into DD).

Model simulation and analysis

Model simulations were carried out on the published models and parameter sets (Locke et al., 2005a,b, 2006) using the Circadian Modelling interface (available from http://www.amillar.org). Simulations were entrained for 20 days under the respective photoperiods and then released into LL or DD for 10 days. Details of the dusk sensitivity analyses are given in the Supplementary information.

Supplementary information

Supplementary information is available at the Molecular Systems Biology website (http://www.nature.com/msb).

Acknowledgements

This work was supported by BBBSRC awards G19886 and E015263 to AJM and by EU FP6 award EUICL0CK to AJM and others. The Centre for Systems Biology at Edinburgh is a Centre for Integrative Systems Biology supported by BBBSRC and EPSRC award D019621.

Author contributions: KDE, KK, PJL and AJM designed the experiments; KDE, KK, AW and PJL performed the experiments; LKB and FN constructed materials; KDE, PEB and DAR designed and
implemented data analysis; KDE and AWT performed data analysis; KDE, AP and OEA performed model simulations; OEA designed and performed model analysis; KDE, OEA and AJM wrote the paper.

Conflict of interest
The authors declare that they have no conflict of interest.

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