Association analysis reveals genetic control underlying natural variation in circadian rhythms in Arabidopsis

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October 1, 2020

Abstract

Circadian clocks have evolved to resonate with external day and night cycles. However, these entrainment signals are not consistent everywhere and vary with latitude, climate and seasonality. This leads to divergent selection for clocks which are locally adapted. To investigate the genetic basis for this circadian variation, we used a Delayed Fluorescence (DF) imaging assay to screen 191 naturally occurring Swedish Arabidopsis accessions for their circadian phenotypes. We demonstrate that period length co-varies with both geography and population sub-structure. Several candidate loci linked to period, phase and Relative Amplitude Error (RAE) were revealed by genome-wide association mapping and candidate genes were investigated using TDNA mutants. We show that natural variation in a single non-synonymous substitution within COR28 is associated with a long-period and late-flowering phenotype similar to that seen in TDNA knock-out mutants. COR28 is a known coordinator of flowering time, freezing tolerance and the circadian clock; all of which may form selective pressure gradients across Sweden. We demonstrate the effect of the COR28-58S SNP in increasing period length through a co-segregation analysis. Finally, we show that period phenotypic tails remain diverged under lower temperatures and follow a distinctive ‘arrow-shaped’ trend indicative of selection for a cold-biased temperature compensation response.

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Funding: This project was supported by the BBSRC via the Earlham institute CSP (BB/P016774/1, AH, HR) and BBSRC Design Future Wheat (BB/P016855/1, AH, RJ).

Abstract: Circadian clocks have evolved to resonate with external day and night cycles. However, these entrainment signals are not consistent everywhere and vary with latitude, climate and seasonality. This leads to divergent selection for clocks which are locally adapted. To investigate the genetic basis for this circadian variation, we used a Delayed Fluorescence (DF) imaging assay to screen 191 naturally occurring Swedish Arabidopsis accessions for their circadian phenotypes. We demonstrate that period length co-varies with both geography and population sub-structure. Several candidate loci linked to period, phase and Relative Amplitude Error (RAE) were revealed by genome-wide association mapping and candidate genes were investigated using TDNA mutants. We show that natural variation in a single non-synonymous substitution within COR28 is associated with a long-period and late-flowering phenotype similar to that seen in TDNA knock-out mutants. COR28 is a known coordinator of flowering time, freezing tolerance and the...
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Keywords: Circadian clock, temperature compensation, Arabidopsis, natural variation, Sweden, 1001 genomes project, GWA-mapping, COR28, ELF3

Acknowledgements: We thank Paul W. Goedhart (Wageningen University) for advice about circular phase statistics using the RCIRCULAR procedure and Grant Calder (John Innes Centre) for his assistance developing the 96-well ROI selection using FIJI. We thank Susan Duncan for help and advice with the flowering time investigation.

We also thank the Dean Research group (JIC) for providing the 191 Swedish accessions and the Liu Research group (Shanghai Institutes for Biological Sciences) for providing the cor28-2, cor27-1 and cor27-1/28-2 mutants. We also thank Ewan Holmes and Georgia Scutter for help with seed harvesting.

Main text:

Introduction

Plants are highly adapted to survive and exploit the daily fluctuations in light and temperature experienced as the earth spins on its axis. The circadian clock plays an intrinsic role in this; integrating temporal cues from the environment to inform photosynthetic, metabolic and developmental processes (Covington et al., 2008; Harmer et al., 2000). A robustly oscillating circadian clock which is highly synchronised to external day-length contributes to the overall fitness of the plant, giving it an edge over competitors, predators and pathogens (Dodd et al., 2005; Green et al., 2002; Ingle et al., 2015; Michael et al., 2003). Circadian rhythms can be quantified by their period (the time taken to complete one cycle), their phase (the time of day in which the rhythm peaks), their amplitude (the change in intensity from a baseline) and their Relative Amplitude Error (RAE) which is the amplitude error divided by the overall amplitude of the rhythm and can be equated to rhythm robustness. In our study, circadian phenotyping was achieved using delayed fluorescence (DF) imaging. Delayed fluorescence levels are circadian regulated and reflect the changes in the photosynthetic state of photosystem II (Gould et al., 2009). DF imaging has been validated as a reliable and flexible tool to measure circadian rhythms in a range of plant models but has not previously been used for phenotyping of a large population of individuals on the scale required for genome-wide association mapping.

Arabidopsis thaliana is a model plant system which has been extensively studied in circadian biology. The Arabidopsis clock is comprised of a series of interlocking negative transcriptional feedback loops connected by key activators that control the oscillation of clock gene expression (Hernando et al., 2017). Key genes within this network include CIRCADIAN CLOCK ASSOCIATED 1 (CCA1) and LATE ELONGATED HYPOCOTYL (LHY) which transcriptionally repress TIMING OF CAB EXPRESSION (TOC1) (Alabadi et al., 2001). Downstream genes interact with the clock to communicate circadian rhythmicity to physiological outputs. Examples include photoperiodic regulation of flowering time and the temporal gating of cold acclimation responses (Kinmonth-Schultz et al., 2013). Flowering under long days is instigated through accumulation of the floral promoter CONSTANS (CO) controlled by the circadian clock component GIGANTEA (Suárez-López et al., 2001). In winter-annuals, flowering is also dependent on the vernalization response of FLOWERING LOCUS C (FLC) and FRIGIDA (FRI) (Bastow et al., 2004; Johanson et al., 2000). Cold tolerance is diurnally activated through alternative splicing of CCA1 which co-regulates the expression of COLD REGULATED (COR) genes alongside light and temperature stress pathways (James et al., 2012; Kinmonth-Schultz et al., 2013).

The endogenous core circadian network is entrained by external stimuli; most notably light and temperature. Day-length, light composition and light intensity have all been shown to affect circadian rhythms (Aschoff, 1979; Más et al., 2003; Pittendrigh & Minis, 1964; Yanovsky & Kay, 2002). Temperature also has a well-documented effect on entrainment of circadian rhythms (Edwards et al., 2005; Gould et al., 2006; Sa-
lome & McClung, 2005) with a degree of period shortening expected under higher temperatures. Circadian rhythms are said to be ‘temperature compensated’; they resist large changes in period-length in response to temperature (Pittendrigh & Harvey, 1954). The extent of temperature compensation has been shown to vary between Arabidopsis accessions (Gould et al., 2006; Kusakina et al., 2014). Rhythm robustness is also strongly affected by temperature, although the temperature which produces the most rhythmic oscillations appears to be dependent on the species and circadian assay used (Kusakina et al., 2014; Rees et al., 2019). Plants with clocks which resonate with environmental conditions are typically fittest, however it has been suggested that in climates with large seasonality there may be a compromise for clocks which are adaptable to rapidly changing day-lengths (Michael et al., 2003).

Natural populations of several important model organisms that exhibit extensive diversity in their circadian behavior have been documented. Pupal eclosion rhythms of Drosophila are latitude dependent; with shorter rhythms, earlier phases and less robust rhythms observed in Northern latitudes (Allemand & David, 1974; Lankinen, 1986). In Arabidopsis, Michel et al. conducted a global study of leaf movement rhythms in 150 accessions and found that day-length of origin country correlated positively with period, but not phase or amplitude. They also identified several loci in the TOC1/PRR family which determined natural variation in period, phase and amplitude independently (Michael et al., 2003). Other investigations have shown allelic diversity of several clock related genes including FLC (Swarup et al., 1999) GI (de Montaigu et al., 2015) and EARLY FLOWERING 3 (ELF3) (Anwer et al., 2014) which contribute to natural circadian phenotypes without fully disrupting the clock mechanism. The positive relationship with period and latitude has also been observed in natural populations of Mimulus guttatus (Greenham et al., 2017) and in cultivated varieties of soybean and tomato (Greenham et al., 2017; Müller et al., 2016), potentially due to unintentional selection for circadian rhythms which function best at particular latitudes.

In this study, we focused on a collection of Arabidopsis accessions from across a large latitudinal range in Sweden; a country with variations in climatic, anthropogenic and day-length factors all possibly influencing clock adaption. Northern latitudes around 63°N have permanent snow cover during winter months and the growing season is much cooler and shorter than in the South. At the solstice, there is almost 3 hours difference in day-length between the North and South. This divergence of climate has led to selection for ecotypes with adapted growth and flowering strategies (Bloomer & Dean, 2017; Shindo et al., 2005). Analysis of the global population structure of Arabidopsis accessions has previously identified Swedish accessions as being genetically distinct from the wider population, with further differentiation within the country between North and South (Horton et al., 2012; Huber et al., 2014; Long et al., 2013; Nordborg et al., 2005). Accessions from South Sweden have high genetic diversity within a relatively small area, perhaps suggesting a historic emigration from central Europe following glacial retreat (Alonso-Blanco et al., 2016). Northern Swedish accessions have lower genetic diversity but larger genome sizes (Long et al., 2013) and also carry a surprising reservoir of drought tolerance genes (Exposito-Alonso et al., 2018). Completion of the 1001 genomes project in 2006 has facilitated a recent expansion in Arabidopsis genome wide association (GWA) studies made possible through the provision of high-quality re-sequenced genotype data. These accessions are publicly available, geo-referenced and genetically inbred making it easy for researchers to perform experiments over several generations under a variety of conditions (Weigel & Mott, 2009). Many of the accessions in this study have also been characterized for other phenotypic traits such as seed dormancy (Kerdaffrec et al., 2016), flowering time (Sasaki et al., 2015), and freezing tolerance (Horton et al., 2016).

The objectives for this research were firstly; to investigate whether circadian diversity exists in Swedish Arabidopsis and to understand to what extent this variation can be explained by either latitude or the underlying population structure. We next wanted to identify and validate genetic variation in candidate genes which might explain the observed circadian variation. To our knowledge, this work provides the first example of a GWA study using circadian phenotypes in plants. Our final objective was to measure circadian rhythms under 10°C and 16°C to test whether divergence of circadian traits persists across a range of temperatures.

Methods
Plant material, plate set-up and growth conditions

Seed for the 192 Arabidopsis accessions were gas sterilised using 3ml hydrochloric acid in 100ml sodium hypochlorite. We then stratified seed in 1ml of sterile water for 2 days at 4°C before plating. Clear 96 well plates with flat bottomed wells (Thermo Scientific, cat no: 10287631) were filled with 250ul of Murashige and Skoog agar media (1.5% Agar, pH5.8, no sucrose). Approximately 20 seeds were added to each well of the plates following a randomized Alpha-lattice design (see Extended Methods) and left to dry until all residual water had evaporated. Each accession was replicated 18 times across the 3 imaging runs. Clear microplate lids (Thermo Scientific, cat no: 10334311) were secured to each plate with micropore tape ensuring that the lid was raised slightly from the top of the wells to allow condensation to circulate freely. Plates were returned to the fridge for 2 more days before transfer to the growth cabinet set at 12:12 L:D cycle at 22°C under approximately 200 μmol m⁻² s⁻¹ white light for 14 days. Prior to the temperature response experiments, accessions from the 191 phenotyping screen were bulked for seed (see Extended Methods). Plates were all grown under 12:12 L:D at 22°C light for 10 days and were then transferred to the temperatures in which they would be imaged for 4 days.

Note: For the temperature and mutant screening experiments, the corner wells of the 96-well plates were not used in the designs as we found that corners had significant effects on circadian rhythms in the 191 dataset; potentially due to these wells drying out more quickly (Supplementary Figure 8).

Imaging conditions

Delayed fluorescence imaging was carried out using Lumo Reteiga CCD cameras (QImaging, Canada) housed in two temperature-controlled cabinets. Six 96-well plates can be imaged simultaneously in each cabinet. Imaging conditions were exactly those described for constant light (L:L) imaging of Brassica leaves in Rees et al. 2019 (Rees et al., 2019). Images were captured every hour with a 1-minute exposure time.

FIJI ROI selection and circadian analysis

Image stacks were imported into FIJI (RRID:SCR_002285) (Schindelin et al., 2012) and regions of interest (ROI) were selected using a semi-automated approach (full details in Extended methods; macro and user guide in Supplementary files 4 and 5). Measurements for integrated density were taken for these regions across the stack using the Multi-measure plugin. The time-series was converted to decimalized time relative to hours elapsed from entrained dawn (T0) and was cropped to 150h to standardize between experiments. Period estimation was done using the online platform Biodare2 (https://biodare2.ed.ac.uk) applying the FFT-NLLS algorithm on a data window of 24-120h with expected periods set to 18-34h. Raw data was baseline and amplitude de-trended prior to analysis. Manual inspection of resulting periods ensured that all arrhythmic traces were excluded from further analysis.

Data analysis and REML

We filtered period estimates from Biodare2 to exclude rhythms with RAE>0.65 and period >28h as we found that rhythms beneath these cut-offs gave the most reliable period estimates. We next used restricted maximum likelihood (REML) to fit a linear mixed model to the 191 accession dataset and thus obtain accession means for period and RAE which were adjusted for the effects of cabinet and experimental run (see details in Extended Methods). An additional step was required to calculate phase means for each accession as the raw phase data is circular and relative to dawn (0) with a full circle representing 24 hours. Phase data was analysed with a circular regression model using the Genstat RCIRCULAR procedure to obtain accession mean phases adjusted for cabinet and run effects (Fisher & Lee, 1992).

For temperature experiments, no period or RAE cut-offs were imposed as we predicted an increase in both these variables with lower temperatures (Dodd et al., 2014; Rees et al., 2019). We fitted linear models to the period and RAE data and identified the contribution of each component by analysis of variance (Supplementary Tables 12-14). The above data analysis was done using Genstat 18th edition (RRID:SCR_014595).
Map figures (Figure 2 and 4) were created using the ggmaps package in R using Google maps (2018) (Kahle & Wickham, 2013).

Principal component analysis

The PCA for genomic variation in the 191 Swedish accessions in Figure 2E was performed using a compressed mixed linear model run in R implementing the GAPIT package (Lipka et al., 2012; Zhang et al., 2010). Genotype data was downloaded from the 1001 genomes dataset filtered for biallelic SNPs with >5% MAF (minor allele frequency). The first two PCs were used to split the accessions into the categories in Figure 2D and F.

GWAS

Genome wide association mapping was conducted using an Accelerated Mixed Model (AMM) via the online web application GWA-Portal (Seren, 2018). Accession means estimated from REML were used for period, phase and RAE with no further transformation necessary. Genotype data used was from the full-sequence 1001 genomes dataset. Results were filtered for the top 10,000 -log10(P) values and then for hits with >5% MAF. Comparison to results from other models confirmed the major peaks and can be viewed in Supplementary Figure 6.

Mutant screening

T-DNA and EMS mutants were obtained from the Nottingham Arabidopsis Stock centre (RRID:SCR_004576) and primers designed using the iSect tool from the SIGnAL website. In addition, seed for cor28-2, cor27-1 and cor28-2/cor27-1 was kindly donated from Hongtao Liu’s laboratory (Shanghai, China) where these mutants have previously been identified as long-period using leaf-movement and ProCCR2:LUC constructs (X. Li et al., 2016). Cor28-2 is a null mutant and cor28-1, cor27-1 and cor27-2 are all knockdown mutants. Gene candidates, mutant IDs and primers used can be viewed in Supplementary File 2. All lines were genotyped to confirm homozygosity by PCR and gel electrophoresis prior to seed bulking other than sco2 which was genotyped by sanger sequencing to reveal the premature stop. We were unable to obtain or confirm homozygous mutants for dnf and fab1c and therefore these lines were not DF screened (Supplementary File 2). SALK lines for elf3 were not obtained as the elf3-1 mutant has been previously confirmed in Box et al. 2015 as having a phenotype of reduced robustness using delayed fluorescence imaging (Box et al., 2015). WT control lines; Col-0 and Ler were grown for seed in parallel to standardize seed quality. For cor28-1, a homozygous WT control was derived by segregation of the original heterozygous TDNA line was used as a control. Five imaging experiments using a single camera and cabinet were carried out with each line present in at least three experiments. We did not apply a mixed model to the mutant screening data because only one camera was used.

Co-segregation analysis

The co-segregation analysis was conducted using a Cleaved Amplified Polymorphic Sequences (CAPS) assay to genotype and DF to phenotype crossed and selfed individual plants.

Primers were designed to amplify a 252bp fragment within COR28 exon 2 (COR28-F : 5’-ACAGTAAGTAAACCAGACC-3’, COR28-R: 5’-TTCTTGACAAAAAAAAACCCCACTC-3’). These fragments were then digested using TaqI (Thermo Fisher Scientific, #ER0671, recognition site: T’CGA). TaqI was added to 10ul of PCR reaction mixture and incubated at 65C for 1.5h according to manufacturer’s instructions. Individuals homozygous for the COR28 SNP 58S (found in the parent St-0) are not recognised by TaqI and so presented as a single 252bp band on an agarose gel (3% Agarose, run for 1.5h at 50V). Individuals homozygous for the COR28 SNP 58W (found in the parent T800) were fully cleaved resulting in two bands 92bp and 160bp in length. Heterozygous individuals therefore had three bands at 252bp, 160bp and 92bp clearly separated by electrophoresis. Restriction digest gels for identifying F2 plants can be viewed in Supplementary file 6. All plants in generations (F0-F1) were vernalised for 12 weeks at 4°C prior to growth at 22°C under long days (16:8h). Parent accessions St-0 and T800 were genotyped to check for homozygosity prior to crossing. All three homozygous F1 individuals were (T800 x St-0) and were also confirmed by the
CAPS assay. These F1 plants were then selfed and the circadian rhythms of the F2 progeny were analysed using DF on detached leaves from 24 day old plants grown under 12:12h light:dark conditions at 22degC. Leaves were placed on 0.5% agar plates and were exposed to 24h constant L:L prior to imaging as described for the 96 well plates.

Results

Quantitative circadian variation across 191 Swedish Arabidopsis accessions

We used delayed fluorescence imaging to characterize circadian rhythms in groups of 14 day old seedlings entrained in 12h L 12h D cycles at 22degC and assayed under free-running conditions of constant light (L:L) at 22degC. Accession means and standard errors (SE) adjusted for experimental effects were obtained using linear mixed models for period, amplitude and RAE and a circular regression model for phase (see Extended Methods). We found a 4.42h difference between the mean period of the fastest (21.28h, SE=0.329) and slowest (25.70h, SE=0.355) accessions tested (Figure 1A). The mean phase of peak DF intensity for each accession occurred over the dark half of the cycle (12-24/0h) with a huge difference (10.32h) from the earliest peaking (12.21h, SE=0.58) to the latest (22.54h, SE=0.20) (Figure 1B). This variation in period and phase is consistent with data previously reported from a global collection of Arabidopsis accessions (Michael et al., 2003). We also measured the robustness of these rhythms by looking at their RAE values. The most rhythmic (lowest RAE) mean was 0.31 (SE=0.023) and the least rhythmic was 0.44 (SE=0.024) representing approximately 20% of the possible range of this trait in this study (Figure 1C). Accession means for circadian phenotypes can be viewed in Supplementary File 1.

To assess the effect of genotype on each trait we performed a likelihood ratio test against a model which omitted genotype as a variance component. Including genotype in the mixed model had highly significant effects on period ($\chi^2 (1 df)=455.57, p<0.0001$) and RAE ($\chi^2 (1)= 65.97, p< 0.0001$). Including genotype in the circular regression model also had a strongly significant effect on phase ($\chi^2 (191)=407, p< 0.0001$). Amplitude was analysed as a log$_{10}$ transformation (see Extended Methods) and there was no significant effect of genotype on Log10Amplitude ($\chi^2 (1)= 0.19, p = 0.6$). With this in mind, we dropped Amplitude as a trait of interest for the temperature and mutant screening experiments. REML output tables (Supplementary Tables 1-3), model checking graphs (Supplementary Figures 1-3) and log-likelihood results (Supplementary Tables 4-7) are available to view.

We observed a strong correlation between the period, RAE and phase of each accession (Supplementary Figure 4), especially between period and phase (Adjusted $R^2 =0.43, p< 0.0001$). We are unaware of any previously published correlation between natural variation in period and RAE. In this study, we found a highly significant negative relationship with longer periods having lower RAE scores (Adjusted $R^2 =0.1, p< 0.0001$).

We chose 10 accessions from the tails of the distributions of mean period, phase and RAE to create “phenotypic tail” groups representing the extremes of each trait. Some accessions represented two or more tail groups and could be split into two master groups of: long period, dusk phased, low RAE accessions and short period, dawn phased, high RAE accessions as explained in Figure 1D. These phenotypic tail groups were used for the temperature experiments described later in this paper (accessions in each tail are listed in Supplementary Tables 8-10).

Our next question was whether other quantified traits co-varied with our circadian data. We used several previously published datasets to examine possible phenotypic relationships between our traits and flowering time (Y. Li et al., 2010; Sasaki et al., 2015), seed dormancy (Kerdaffrec et al., 2016) and freezing tolerance (Horton et al., 2016) (Supplementary Table 11). Only flowering time data from Li et al. (2010) showed a moderate correlation with period ($p<0.01, R^2 =0.28$) and phase ($p<0.01, R^2 =0.26$). This flowering time data is based on flowering times recorded in conditions of variable day-length reflecting annual conditions of a natural Swedish or Spanish climate (Y. Li et al., 2010). However, no significant correlations were found between flowering times recorded under long-days under either 10°C or 16°C by Horton et al. (2016), indicating that the relationship is highly dependent on the exact experimental conditions used.
We then investigated whether circadian traits varied according to the original location of the accessions. Period was found to be significantly correlated with longitude and latitude ($p < 0.001$), however a considerable amount of variation remained unexplained (Adjusted $R^2 < 0.1$) (Supplementary Figure 5). Despite this weak overall association with latitude, we observed an obvious segregation of accessions from the phenotypic tails of the period distribution, with the shortest period accessions in Northern and mid-latitude regions and the longest period accessions in the south (Figure 2A). No significant correlation with altitude was observed for these accessions.

Following on from this, we asked the question whether population sub-structure could better explain the distribution of period phenotypes. We re-classified the population into three groups split across the axes of the first two principal components (which together explain over 15% of the total genetic variation) (Figure 2D and E). PC.B broadly contained Northern accessions (N=54, red points in Figure 2D). The Southern accessions split into two sub-groups existing within the same geographic region; PC.A (N=111) and PC.C (N=26). We found that period is significantly longer in group PC.C (Figure 2F) ($F(2,188)=8.39$, TukeyHSD for PC.A and PC.C $p< 0.01$, One-way ANOVA). PC.A had mean periods similar to those obtained from Northern accessions (PC.B, Figure 2F). PC.C accessions also have significantly earlier phase peaks than those in PC.A (One-way ANOVA, $F(2,188)=6.9149$, TukeyHSD for PC.A and PC.C $p< 0.001$).

We removed the effects of PC groups from period length variation and found that the effect of latitude on the remaining period variation was much less significant (REML Model: Period $\sim (1 | PC.blocks) + \text{Latitude}$, applying Satterthwaite’s method. $F(1,7.23) = 8.87$, $p > 0.01$). This indicates that period variation with latitude is due to the different locations of sub-populations, rather than due to adaption following a latitudinal cline. Accessions in the long period tail group described in Figure 1A belonged to both PC.A (N=6) and PC.C (N=4), suggesting that long-period traits may have been preserved in both sub-populations in this area.

**GWA mapping reveals loci associated with natural circadian variation**

We used the online GWA-portal to perform association mapping on the three circadian traits found to be significantly affected by genetic variation (Seren, 2018). We used an Accelerated Mixed Model (AMM) to account for population structure, although the other association models (linear and non-parametric) gave similar results (See Supplementary Figure 6). Pseudo-heritability estimates for each trait were given as an output of the GWA analysis: period=71%, RAE=37% and phase=13%.

GWA identified multiple genomic regions associated with period, phase and RAE as shown in Figure 3. We used $-\log_{10}(p) \geq 6.5$ as an arbitrary p-score cut-off to select SNPs for further investigation. This threshold is conservative compared to several other previously published studies in *Arabidopsis* (Kooke et al., 2016; Proietti et al., 2018; van Rooijen et al., 2015). We investigated known core circadian and flowering time genes to see whether these were significantly associated with any of the traits measured. Other than *ELF3* (see below), none of these genes fell within the window of association for significant SNPs. Genes in regions 30kb upstream or downstream of the most significant SNPs were considered as potential gene candidates and were selected for further analysis based on their previously attributed functionality and designated GO term. Genes involved in circadian rhythms, flowering time or chloroplast regulation were given prevalence (details in Supplementary File 2). The most significant associations had three SNPs with a $-\log_{10}(p)$ score of 10.4-11.5, found on chromosome 4 associated with period variation. Within this interval we identified a non-synonymous SNP in the gene *COLD-REGULATED GENE 28*, a gene that has previously been identified as a negative regulator of several core clock genes (*PRR7, TOC1, PRR5* and *ELF4*) and is also implicated in the trade-off between flowering-time and freezing tolerance (X. Li et al., 2016; Wang et al., 2017). The substitution resulted in a tryptophan (W) to serine (S) amino acid change at position 58 within the second exon of *COR28* (Figure 4A). This had a SIFT score of 0 indicating a highly probable deleterious effect on protein function. 16 accessions in this study had the minor allele, all found in the South of Sweden. The 58S accessions had a period 1.29h longer than the 58W accessions ($t(17.4)=-7.46$, $p< 0.001$, Welch Two Sample t-test) (see Figure 4B) and mostly belonged to the genomic sub-group PC.A. The long-period of cluster PC.C therefore cannot be explained by variation in this SNP. We used the online tool Polymorph 1001 to
look for other variants with the serine substitution and found only 5 other variants not assayed in this study (making 21 in total), all of which were also from the South of Sweden (Figure 4C).

We used flowering time data by Li et al. 2010 (Y. Li et al., 2010) and Sasaki et al. 2016 (Sasaki et al., 2015) to look for flowering time differences between the accessions carrying the COR28 -58W major allele and the -58S minor allele (see Table 1). We found that the accessions with COR28 -58S had significantly extended flowering times under simulated seasons for Sweden and Spain (Li) and under a 10°C and 16°C long-day temperature regime (Sasaki) (see Figure 4D). This complements previous findings that COR28 is a flowering promoter and that modifications to this gene increase flowering time as well as lengthening period.

Other significant SNPs associated with period (chromosome 3) and phase (chromosome 1) identified gene candidates involved in chloroplast function (PARC6, SCO2, DNF, AT1G71015 and FAB1C). The DF circadian output we used to assay these traits is based on oscillating activation phases of PSII (Goltsev V, Zaharieva I, Chernev P, 2009; Jursinic, 1986) and therefore natural variation in genes regulating chloroplast functionally could also affect the DF output in these candidates.

A RAE associated SNP on chromosome 2 was found 5726bp upstream from a SNP in ELF3 previously characterized as the ELF3-Sha allele. Our associated SNP was found to be under strong linkage disequilibrium with the ELF3-Sha allele ($R^2=0.86$). The alanine-to-valine transition in amino acid position 362 has been associated with naturally occurring alternations to periodicity and robustness in accessions from Central Asia, specifically Tajikistan (Anwer et al., 2014). Here, 13 Swedish accessions were shown to carry the ELF-Sha allele and had mean RAE ratios 0.032 higher on average than for the other accessions. We substantiate evidence that ELF3-Sha accessions have lower rhythmicity and extend the global range of this allele into Northern Sweden. A surprising kinship between Arabidopsis accessions from Northern Sweden and Central Asia has been previously demonstrated through analysis of global population structure and indicates that the presence of this allele has not evolved convergently between the two populations (Nordborg et al., 2005).

Validation of COR28 effects using knockout lines and co-segregation analysis
11 mutant lines representing eight gene candidates were genotyped to confirm homozygous mutations before being bulked for seed. Confirmed mutants were then assayed for their circadian rhythms using DF under the same conditions as for the 191 accession screen. Mutant details and genotyping results can be viewed in Supplementary File 2. Cor27 mutants and the double mutant cor27-1/28-2 were also assayed ascor27 is known to be partially redundant with cor28. Analysis of the DF rhythms confirmed that periods in cor28 mutants (cor28-1 =25.3h, SD=1.73; cor28-2 =25.3h, SD=1.78) were significantly longer than their respective controls (23.9h, SD=1.62 and 24.0h SD=2.02) (Welch Two Sample t-test $p<0.05$) as shown in Table 2. This confirms results previously reported using leaf movement, qPCR and ProCCR2:LUC bioluminescence rhythms (X. Li et al., 2016; Wang et al., 2017). Cor27 mutants showed no significant difference in period length, however the double knock-out cor27-1/28-2 had an exaggerated long period phenotype (26.4h, SD=2.37). The peak phases of cor27, cor28 and especially cor27-1/28-2 were also earlier than for Col-0, with cor27-1/28-2 having a mean phase peak 2.6h earlier than the control. Other mutants characterized were not found to have significantly altered periods or phases compared to their WT controls (Supplementary Tables 12-14).

Finally, we carried out a co-segregation analysis for the COR2858S allele to demonstrate that this SNP was causal for the long period phenotype. Crosses were made between the line T800 (Watkins 6133) and St-0 (Watkins 8387), with T800 carrying the 58S minor allele and St-0 carrying the 58W major allele. Parent individuals were genotyped using a CAPs assay prior to crossing and were found to be homozygous for their respective SNPs. We selfed three F1 individuals from this cross and analyzed the genotypes and circadian period phenotypes in the F2 offspring using DF on dissected leaves (Table 3.) Restriction digest gels from F2 genotyping can be viewed in Supplementary File 6. F2 individuals carrying the 58S SNP had a mean
period length of 23.7h (SD=1.95) compared to a mean of 21.3h (SD=1.65) for those carrying the 58W SNP (a difference of 2.4h) which was shown to be highly significant (F(2,62)=12.27, p<0.0001, One-way ANOVA. TukeyHSD for SNP 58S and 58W: p<0.001). Heterozygous individuals had period lengths which were on average closer to the long period phenotypes of the 58S homozygote individuals (23.1h (SD=1.95)) suggesting a semi-dominant effect for the 58S allele.

Table 3

The phenotypic tails of circadian traits remain largely diverged under lower temperatures

The variation characterized in the first part of this paper was observed at 22°C which was chosen to make the data comparable to several previously published studies of interest. We wanted to test whether the circadian phenotypic diversity we observed at 22°C continued at temperatures of 16°C and 10°C; closer to those found in the natural Swedish environment. To simplify the dataset, we selected 10 accessions to represent each of the six phenotypic tails as highlighted in Figure 1. We wished to investigate whether reduced temperatures would affect the phenotypic tails equally, drive them further apart or lead to a convergence of their phenotypes. Our null hypothesis was that the phenotypic diversity seen at 22°C would exist consistently at lower temperatures with no differential effect on the phenotypic tail groups.

The results show that decreasing temperature had a massive overall effect on all three circadian outputs, particularly RAE and phase (accession means can be viewed in Supplementary File 3). Overall, the divergence between the tail groups was largely maintained, although the gap reduced at 10°C (Figure 5a-c). For each trait, a general linear model was fitted to the data in order to test the significance of explanatory factors (Supplementary Tables 15-17).

For period, membership to the short or long tail groups was the largest explanatory variable and the group means remained clearly distinct across all temperatures. Temperature also had a large overall effect on period, with rhythms at 16°C running much slower than at 22°C. At 10°C periods were again shorter, accompanied by higher RAE (reduced rhythm robustness) (Supplementary Figure 7). The difference in the period temperature responses of the two groups can be seen by the gradients of the thick colored lines in Figure 5A. There was also significant variation between the temperature response of individual accessions within each tail group, especially in the long period group (see thin grey lines in Figure 5A). Interestingly, Col-0 reacted very differently to the Swedish accessions tested, showing an almost linear decrease in period with increasing temperature (dashed line in 5A).

For RAE, which we equate to rhythm robustness, we found that temperature had an even greater effect than for period, with rhythms at 10°C showing a marked decrease in robustness (Figure 5B). The tails converge as the temperature decreases, with lower temperatures having an especially large effect on the low RAE group.

For phase, decreasing temperature to 10°C caused a large shift of approximately 7.4 hours towards dawn accompanied by increased variability for each accession (Figure 5C). The means of the two phase-tail groups remained distinct across the three temperatures and there was no significant difference in their relative change of phase with temperature.

We also conducted an independent experiment with the phase tail accessions to verify the phase estimates from the two seed batches measured at 22°C (Supplementary Figure 9).

Across all traits we observed an increase in the within-accession variability at lower temperatures indicated by larger standard deviations in period, higher RAE scores and a greater number of rhythms being rejected from Biodare2 analysis as arrhythmic.

Discussion

We measured DF rhythms in 191 naturally occurring Swedish Arabidopsis accessions and show that circadian phenotypes display considerable variation. This variation does not conform to previously described latitudinal clines in Arabidopsis (Michael et al., 2003), as we find the longest periods in accessions from...
the South of the country and the shortest periods in the North. In our study, circadian phenotypes are also clustered into geographical groups rather than following a distinct latitudinal cline. We suggest that the period variation we observe in the Swedish population is due largely to founder effects from ancestral migrations of individuals adapted to different selection pressures (Flohr et al., 2013), and may be influenced by serendipitous fixing of alleles through genetic drift (Song et al., 2006).

We found a high level of correlation between circadian period, phase and RAE in these accessions; period was negatively correlated with both RAE and phase. Interestingly, in the study by Michael et al, period, phase and amplitude were reported to vary independently for leaf movement rhythms in a global panel (Michael et al., 2003).

Periods were found to be significantly different between two sub-populations co-existing in the same geographical area in the South. In spite of this, accessions with very long periods (belonging to the long-period phenotypic tail) were found in both of these populations and the minor COR28 SNP was also found in both the PCA and PC.C sub-groups.

We show that circadian variation assayed by DF is genetically heritable and is associated with several highly significant polymorphisms, two of which (ELF3 and COR28) had previously acknowledged circadian functions. COR28 is partially redundant with its partner COR27 and acts both upstream and downstream of the circadian clock (X. Li et al., 2016; Wang et al., 2017). TDNA insertions in these genes have been shown to lengthen periods, extend flowering time and increase freezing tolerance (X. Li et al., 2016; Wang et al., 2017). Natural allelic diversity in COR28 has not previously been described. Here, we show that a set of 16 naturally occurring accessions from southern Sweden have a W58S amino acid substitution which results in a long period comparable to that seen in cor28 TDNA insert mutants. COR28 and COR27 are expressed in a blue-light and temperature dependent manner, plausibly suggesting why this gene could be under selection in the Swedish environment. The mechanism through which W58S affects the function of COR28 remains unclear. COR28 is a small peptide of ~26kDa and so may not require active transport for nuclear localization (Hicks GR, 2013). No DNA binding domains have been identified in the COR28 sequence, however it has been suggested to regulate TOC1 and PRR5 transcription through the formation of protein complexes (X. Li et al., 2016). It is possible that this modification affects the ability of COR28 to form these protein interactions.

We also identified ELF-sha alleles in Northern accessions which were associated with higher RAE ratios. In barley, a mutant ortholog of ELF3; ema8 was shown to have been positively selected for growth in high-latitude environments, particularly in Scandinavia (Zakhrabekova et al., 2012). ema8 cultivars are rapid flowering enabling survival under short growing seasons, but also have severely attenuated circadian function (Faure et al., 2012).

Finally, we investigated the effect of temperature on natural circadian variation between accessions with divergent circadian phenotypes. Temperature had a large effect on period and an even greater one on phase and RAE means in both tail groups, indicating a low level of temperature compensation for these outputs. This also shows that the forces governing compensation for period do not act equally on maintaining constant RAE or peak phase. Interestingly, rhythms appeared to be most robust at 22degC, which might not be expected given that the warmest months in Sweden average around 15-17degC. A similar loss of rhythm robustness at lower temperatures has been observed in wheat (Rees et al., 2019).

Period had a non-linear relationship with temperature in these accessions, lengthening from 22degC to 16degC, before shortening again at 10C. This arrow shaped profile likely reflects two interacting forces at work; 1) between 22degC and 27degC the acceleration of rhythms due to increased rate kinetics and 2) between 22degC and 10degC the balancing forces of circadian temperature compensation. Gould et al. found a similar effect of temperature on period using leaf-movement rhythms at 22degC, 17degC and 12degC, and showed that the cold temperature compensation response works through an independent mechanism to the hot temperature compensation response (Gould et al., 2006). The profile suggests that temperature compensation is biased towards correction at colder temperatures in these accessions. It is possible that adaptation
to a cold climate has selected for a cold compensation response to overcome excessive deceleration of the clock, although we are unable to explain why the rhythms should be even shorter at 10degC than at 16degC or why the shortening of periods are accompanied by a loss of overall rhythmicity. Although divergence between the phenotypic tails decreased at lower temperatures, the groups remained largely separate, reconfir

Authors' contributions:

This project was conceptualized by AH and HR. HR designed and conducted imaging experiments and mutant screening. JKMB and HR carried out statistical analysis and data processing. The GWA analysis was done by RJ and HR. All authors contributed to interpretation of results. The paper was written by HR with assistance from RJ, JKMB and AH. All authors approved the final manuscript.


patterns of genetic variation in worldwide Arabidopsis thaliana accessions from the RegMap panel. *Nature Genetics*, 44 (2), 212–216. https://doi.org/10.1038/ng.1042


<table>
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<tr>
<th>Datasets</th>
<th>Days to flower</th>
<th>Welch Two sample t-test</th>
<th>Welch Two sample t-test</th>
<th>Welch Two sample t-test</th>
</tr>
</thead>
<tbody>
<tr>
<td>Li (49)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Sweden Average</td>
<td>N for 58W</td>
<td>Mean 58W</td>
<td>Mean 58S</td>
<td>df</td>
</tr>
<tr>
<td>Average</td>
<td>10</td>
<td>54.43</td>
<td>143.66</td>
<td>10.8</td>
</tr>
</tbody>
</table>

In 16°C, 1145 18 83.24 100.92 17.8 4.93 <0.001

Sasaki (47) 16°C 73.97 94.91 17.6 5.75 <0.001
Table 1. Longer flowering time correlation with COR28W58S Allele.

N=Number of accessions used for each COR28 allele from a global population. Flowering data means reflect an average of all replicates within each study by Li et al. 2010 or Sasaki et al. 2016.

<table>
<thead>
<tr>
<th>Plant ID</th>
<th>N</th>
<th>Period mean (hours)</th>
<th>Period SD</th>
<th>RAE mean</th>
<th>RAE SD</th>
<th>Phase circular mean (hours)</th>
</tr>
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<tbody>
<tr>
<td>col-0</td>
<td>180</td>
<td>24</td>
<td>2.02</td>
<td>0.42</td>
<td>0.18</td>
<td>15.35</td>
</tr>
<tr>
<td>cor27-1</td>
<td>108</td>
<td>24</td>
<td>1.75</td>
<td>0.34</td>
<td>0.18</td>
<td>13.39</td>
</tr>
<tr>
<td>cor27-2</td>
<td>107</td>
<td>24.2</td>
<td>2.21</td>
<td>0.29</td>
<td>0.14</td>
<td>14.02</td>
</tr>
<tr>
<td>cor28-2</td>
<td>158</td>
<td>25.3</td>
<td>1.78</td>
<td>0.28</td>
<td>0.13</td>
<td>12.93</td>
</tr>
<tr>
<td>cor28-2/27-1</td>
<td>101</td>
<td>26.5</td>
<td>2.33</td>
<td>0.28</td>
<td>0.14</td>
<td>12.67</td>
</tr>
<tr>
<td>mybl2</td>
<td>103</td>
<td>24</td>
<td>1.97</td>
<td>0.37</td>
<td>0.17</td>
<td>14.27</td>
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<tr>
<td>parc6</td>
<td>97</td>
<td>23.9</td>
<td>2.08</td>
<td>0.37</td>
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<td>13.91</td>
</tr>
<tr>
<td>parc6-1</td>
<td>104</td>
<td>23.4</td>
<td>1.77</td>
<td>0.35</td>
<td>0.17</td>
<td>14.95</td>
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<tr>
<td>atg1g71015</td>
<td>102</td>
<td>23.7</td>
<td>2.11</td>
<td>0.38</td>
<td>0.16</td>
<td>14.45</td>
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<tr>
<td>WT for cor28-1</td>
<td>72</td>
<td>23.9</td>
<td>1.62</td>
<td>0.33</td>
<td>0.17</td>
<td>14.54</td>
</tr>
<tr>
<td>cor28-1</td>
<td>70</td>
<td>25.3</td>
<td>1.73</td>
<td>0.28</td>
<td>0.16</td>
<td>12.95</td>
</tr>
<tr>
<td>ler</td>
<td>108</td>
<td>23.2</td>
<td>2.55</td>
<td>0.43</td>
<td>0.17</td>
<td>16.37</td>
</tr>
<tr>
<td>sco2</td>
<td>101</td>
<td>23.7</td>
<td>1.95</td>
<td>0.51</td>
<td>0.16</td>
<td>15.33</td>
</tr>
</tbody>
</table>

Table 2. Mean results from DF screening of GWA candidate mutant lines.

Mutants are listed under their respective WT controls highlighted in grey. Circular means for phase were calculated using the ‘circular’ package in R.

<table>
<thead>
<tr>
<th>F1 ID</th>
<th>Genotype</th>
<th>N plants genotyped</th>
<th>Expected df Ratio of genotypes</th>
<th>X^2</th>
<th>p-value</th>
<th>Mean Period (h)</th>
<th>SD Period</th>
<th>F-value</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>T1A1</td>
<td>HZ</td>
<td>9</td>
<td>10.5</td>
<td>2</td>
<td>0.43</td>
<td>22.09</td>
<td>1.139</td>
<td>26.55</td>
<td>&lt;0.00001 ***</td>
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<tr>
<td></td>
<td>HM</td>
<td>6</td>
<td>5.25</td>
<td>2</td>
<td>0.81</td>
<td>24.38</td>
<td>0.902</td>
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</tr>
<tr>
<td></td>
<td>58S</td>
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<td>2</td>
<td>0.902</td>
<td>20.50</td>
<td>0.469</td>
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<td></td>
</tr>
<tr>
<td>T1A2</td>
<td>HZ</td>
<td>17</td>
<td>13</td>
<td>2</td>
<td>3.15</td>
<td>23.75</td>
<td>1.92</td>
<td>1.826</td>
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</tr>
<tr>
<td></td>
<td>HM</td>
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<td>6.5</td>
<td>2</td>
<td>0.20</td>
<td>22.51</td>
<td>3.33</td>
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<tr>
<td></td>
<td>58S</td>
<td>6</td>
<td>6.5</td>
<td>2</td>
<td>0.184</td>
<td>21.96</td>
<td>1.92</td>
<td></td>
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<tr>
<td>T1A3</td>
<td>HZ</td>
<td>14</td>
<td>13.5</td>
<td>2</td>
<td>3.67</td>
<td>23.05</td>
<td>1.043</td>
<td>4.08</td>
<td>0.0299 *</td>
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<tr>
<td></td>
<td>HM</td>
<td>3</td>
<td>6.75</td>
<td>2</td>
<td>0.16</td>
<td>23.58</td>
<td>2.050</td>
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</tr>
<tr>
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<td>58S</td>
<td>10</td>
<td>6.75</td>
<td>2</td>
<td>0.0299</td>
<td>21.47</td>
<td>1.856</td>
<td></td>
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</tr>
</tbody>
</table>
Table 3. Results from co-segregation analysis of the \textit{COR28} SNP 58S with long period.

Three heterozygous F1 plants were selfed to produce F2 offspring broadly following a Mendelian segregation ratio of 2:1:1 tested using a Chi-Square test. F2 individuals were phenotyped using DF imaging on detached leaves and were analysed within each F1 group using a One-way ANOVA (analysis across all F1 individuals presented in main text).

Figure 1- Circadian variation in 191 naturally occurring accessions from Sweden

Delayed fluorescence rhythms were characterised by period, phase and RAE and show significant variation (A-C). Colours represent the 10 most extreme accessions for each phenotype which we use as tail groups for further analysis. Some accessions belong to multiple tail groups as is shown in D. This reflects the strong correlation between circadian characteristics (See Supplementary Figure 4). Number of individual wells contributing to the mean of each accession ranged from 4 to 18, with each well representing the mean rhythms of approximately 15 seedlings.

Figure 2- Period phenotypes are linked by both the geography and the genetic relatedness of accessions across Sweden

Figure A shows the location of origin of the 10 accessions with the shortest (red triangles) and longest (blue triangles) periods. The longest periods are found near the city of Malmo where day length and temperature fluctuates less throughout the year than in Ornskoldsvik where periods tend to be shorter (B,C). Day-length and temperature averages were downloaded from timeanddate.com and are based on predictions for 2019 (Thorsen, 1995).

PC analysis revealed a sub-clade within the Southern Swedish accessions; PC.C coloured light blue in plots D-F. This was used to distinguish accessions with significantly longer period phenotype than in the other PC groups; PC.A which represents the remaining southern accessions in yellow and PC.B which represents the northern accessions in red. Map figures were created using the ggrepas package in R using Google maps (accessed 2018) (Kahle & Wickham, 2013).

Figure 3- GWA analysis reveals candidate genes underlying natural variation in DF circadian phenotypes

Manhattan plots show AMM derived SNP associations with period (top panel), phase (middle panel) and RAE (bottom panel). Data shown has >5% MAF and reflects the 10,000 highest –log10(p) values. Gene candidates selected for further analysis are labelled in red with arrows. The red dotted line indicates the 6.5 –log10(p) arbitrary threshold used for this study. Vertical grey lines are labelled under the x axis and represent positions of known circadian and flowering time genes with altered circadian phenotypes previously reported in mutants.

Figure 4- Natural Variation in \textit{COR28} is associated with covariation in period and flowering time

The SNP in \textit{COR28} associated with period variation causes a tryptophan (W) to serine (S) amino acid change within the second exon (A). Accessions carrying the 58S allele had periods 1.29h longer on average than those with the 58W allele based on the 191-accession data from this study (B). Out of the 191 accessions originally phenotyped, 16 had the 58S allele (red triangular points). Globally, 5 further 58S accessions were identified (orange triangles) and all were found in the south of Sweden (C). Accessions carrying the 58S allele also had flowering times delayed by 89.23 days compared to those carrying the 58W allele (data shown is based on flowering times from plants grown under Swedish conditions previously published in Li et al. 2010 (D).

Figure 5- Effects of temperature on circadian rhythms in phenotypic tails

We selected 10 accessions to represent the phenotypic tails of each trait as highlighted in Figure 1. Period, phase and RAE in these accessions were then re-quantified at 10°C, 16°C and 22°C to see if the tails remained
diverged at lower temperatures. In Figure A, light grey lines show the mean periods of accessions in the long period tail group with the thick blue line showing the mean period for the whole group. Black lines show the mean periods for accessions in the short period group with the thick red line reflecting the mean for the whole short period group. In Figure B, light grey lines show the mean periods of accessions with high RAE (low robustness) with the thick maroon line indicating the overall high RAE tail mean. The black lines show the mean periods for accessions with low RAE (high robustness). The green line shows the overall low RAE tail mean. In both Figure A and B the dotted line is the mean for Col-0 at each temperature. Figure C shows individual phase estimates from accessions belonging to the dusk-phased (dark filled point) or dawn-phased (open point) tail groups plotted as clock plots relative to 24/0 representing dawn. Tail group phase means are indicated by coloured arrows (dusk tail in orange and dawn tail in purple). Sample size contributing to accession mean at each temperature ranged from 14 to 24.)
Ornsköldsvik
Long Period tail (10 individuals)
Remaining Accessions
(171 individuals)
Short Period tail (10 individuals)
Malmö
PC.A (111 individuals)
PC.B (54 individuals)
PC.C (26 individuals)
<table>
<thead>
<tr>
<th>Chromosome</th>
<th>Trait</th>
<th>LHY</th>
<th>STIPL1</th>
<th>GI</th>
<th>SKIP</th>
<th>ELF3</th>
<th>ELF4</th>
<th>PRR9, CCA1</th>
<th>RVE8</th>
<th>LUX</th>
<th>FRI</th>
<th>PRMT5</th>
<th>PRR7</th>
<th>FLC</th>
<th>PRR5</th>
<th>ZTL, BOA, PRR3, TOC1</th>
<th>Cor27</th>
<th>COR28</th>
<th>PARC6, SCO2, DNF</th>
<th>MYBL2</th>
<th>AT1G71015</th>
<th>FAB1C</th>
<th>ELF3</th>
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</thead>
<tbody>
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</tbody>
</table>

### Diagram

The diagram shows the relationship between various traits and their associated positions on different chromosomes. Each chromosome is labeled, and specific traits are highlighted with corresponding log10 P-values. The traits include LHY, STIPL1, GI, SKIP, ELF3, ELF4, PRR9, CCA1, RVE8, LUX, FRI, PRMT5, PRR7, FLC, PRR5, ZTL, BOA, PRR3, TOC1, Cor27, COR28, PARC6, SCO2, and DNF. The diagram also highlights the position of MYBL2 and AT1G71015, as well as FAB1C and ELF3, which are associated with specific P-values.
Cphenotyped

Gnot in our 191 screen

Gphenotyped

A

Chr 4
Reverse Strand

4:16284072
tGg/tCg
W58S

162830001628400016285000
ATG TAA
Chr 4
Reverse Strand

B

Period

58W 58S

D

Mean days to flower

58W 58S

58W

58S

58W

58S