

# Differences in leaf and wood traits predict phenological sensitivity to daylength more than temperature

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## Research questions

1. Do phenological cue-trait relationships change across transects? Populations?
2. How do budburst cues relate to functional traits in temperate woody species?
3. How do shrub and tree species differ in their cue-trait relationships?

## Materials and Methods

### Field sampling

We combined *in situ* trait data with budburst data from two growth chamber cutting experiments conducted across eastern and western deciduous forests in North America. Both suites of data were collected from populations that span latitudinal gradient of 4-6° for eastern and western communities respectively. Our trait measurements were taken across eight populations, of which there were four eastern populations—Harvard Forest, Massachusetts, USA (42.55°N, 72.20°W), White Mountains, New Hampshire, USA (44.11°N, 52.14°W), Second College Grant, New Hampshire, USA (44.79°N, 50.66°W), and St. Hippolyte, Quebec, Canada (45.98°N, 74.01°W) and four western population—E.C. Manning Park (49.06°N, 120.78°W), Sun Peaks (50.88°N, 119.89°W), Alex Fraser Research Forest (52.14°N, 122.14°W), and Smithers (54.78°N, 127.17°W), British Columbia (BC), Canada. For the growth chamber studies, cuttings were collected from the most southern and northern populations in each transect ( $n_{pop}=4$ ).

Across all eight populations, we measured a diverse assemblage of species from the understory and canopy layers. We selected the angiosperm species that were most abundant in our forest communities, with an aim to maximize the number of closely related species and congeners between our eastern and western communities. We focused on angiosperm species only, as these species are more likely to have similar environmental controls to their leafout phenology, and to standardize our leaf trait comparisons by excluding gymnosperm species with.

## Functional traits

We measured all traits in the summer prior to each growth chamber study. For our eastern transect, traits were measured from June 8 to June 25, 2015, and from May 29 to July 30, 2019 for our western transect. At each population, we measured a total of five traits: height, diameter at breast height (DBH), stem specific density (SSD), leaf mass area (LMA), and the percent leaf nitrogen content (LNC). Each trait was measured for each species present at a populations, resulting in us measuring 1-10 healthy, adult, individuals per species at each population. We also used the WSL xylem database (Schweingruber and Landolt, 2010) to collected data on species ring-type for the 72.3% of our species represented in the database.

We measured traits in accordance to the methods discusseed by Pérez-Harguindeguy et al. (2013). We calculated tree height using trigonometric methods, using a TruePulse 200L rangefinder and measured DBH 1.42m from the ground. For shrub heights, we measured the distance from the ground to the height of the top foliage and measured stem diameters at approximately 1 cm above ground-level. To measure stem specific density, we collected a 10cm sample of branch wood, taken close to the base of the stem. All samples were kept cool during transport and measurements of stem volume taken within 12 hours of sample collection. We measured wood volume using the water displacement method. We dried our stem samples upon returning from the field at 105°C for 24h, and weighed their dry mass. Stem specific density was calculated as the dry mass of the sample over its fresh volume.

For our two leaf traits, we haphazardly selected five, fully expanded, and hardened leaves, avoiding leaves with considerable herbivory damage. We kept leaves cool during sampling and transport. For each leaf, we took high resolution scans using a Canon flatbed scanner (CanoScan Lide 220) within 12 hours of collection. We estimated leaf area using the ImageJ software (version X). Upon returning from the field, we dried our leaves for 48h at 70° and weighed each leaf using a precision balance. Leaf mass area was calculated as the ratio of the leaf dry mass over it's area.

## Growth chamber study

For our growth chamber study, we collected branch cuttings from our highest and lowest latitude populations in each transect. In both our eastern and western controlled environment study, we included two temperature treatments and daylength, for a total of eight distinct treatments. Our treatments included two levels of chilling—with our eastern study having no additional chilling or 30 days at 4°C, and our western study 30 days or 70 days of chilling at 4°C, with all non-field chilling occurring under dark conditions. Our forcing treatments included either a cool regime of 15:5°C or a warm regime of 20:10°C, and our photoperiod treatments consisted of of either 8 or 12 hour daylengths. We recorded budburst stages of each sample every 1-3 days for up to four months, defining the day of budburst as the day of budbreak or shoot elongation (defined as code 07 by (?)). For a more detailed discussion of study sample collection and methods see Loughnan and Wolkovich (In prep).

## Statistical Anlaysis

Our analysis combined our *in situ* trait data with budburst data from the growth chamber studies. For each trait, we developed a joint Bayesian model, in which the relationship between traits and cues is used to estimate budburst. This statistical approach improves upon previous analyses of multiple traits, as it allows us to carry through uncertainty between trait and phenology data—allowing us to better partition the drivers of variation in species phenologies

Our joint model consists of two parts. The first is a hierarchical linear model, in which we have partitioned the variation of individual observations ( $i$ ) of a given trait value ( $y_{\text{trait}[i]}$ ) to account for the

effects of species (*sp id*), population-level differences arising from transects (*transect id*), as well as the interaction between transects and latitude (*transect*  $\times$  *latitude*), and finally, residual variation or ‘measurement error’ ( $\sigma_{\text{trait}}$ ).

$$\begin{aligned}\mu_{\text{trait}} &= \alpha_{\text{grand trait}} + \alpha_{\text{sp}[\text{sp id}]} + \beta_{\text{transect}} \times \text{transect} + \beta_{\text{transect} \times \text{latitude}} \times (\text{transect} \times \text{latitude}) \\ \alpha_{\text{sp}} &\sim \text{normal}(0, \sigma_{\text{sp}}) \\ y_{\text{trait}} &\sim \text{normal}(\mu_{\text{trait}}, \sigma_{\text{trait}})\end{aligned}\quad (1)$$

Transect was included as a categorical variable and latitude as a continuous variable in our model. Traits were modeled using natural units, with the exception of LMA, which was rescaled by 100. While each of our traits collected at the individual level were run using separate models, comparisons across species ring-porosity were made using the posterior estimates of our height model. This allowed us to best account for inherent differences in wood anatomy across species and growth form.

We used partial pooling to uniquely estimate species-level variance ( $\alpha_{\text{sp}[\text{sp id}]}$ )—which controls for variation in the number of trait estimates per species and trait variability—which are then used as predictors of species-level estimates of each cue ( $\beta_{\text{chill}[\text{sp}]}$ ,  $\beta_{\text{force}[\text{sp}]}$ ,  $\beta_{\text{photo}[\text{sp}]}$ ).

$$\begin{aligned}\beta_{\text{chill}[\text{sp}]} &= \alpha_{\text{chill}[\text{sp}]} + \beta_{\text{trait.chill}} \times \alpha_{\text{trait sp}[\text{sp}]} \\ \beta_{\text{force}[\text{sp}]} &= \alpha_{\text{force}[\text{sp}]} + \beta_{\text{trait.force}} \times \alpha_{\text{trait sp}[\text{sp}]} \\ \beta_{\text{photo}[\text{sp}]} &= \alpha_{\text{photo}[\text{sp}]} + \beta_{\text{trait.photo}} \times \alpha_{\text{trait sp}[\text{sp}]}\end{aligned}\quad (2)$$

In addition to the species-level estimates, the second part of our model estimates the overall effect of each trait on each cue ( $\beta_{\text{trait.chill}}$ ,  $\beta_{\text{trait.force}}$ ,  $\beta_{\text{trait.photo}}$ ). From this we can estimate how well trait effects explain species-level differences by estimating the the species-level cue variation not explained by traits ( $\alpha_{\text{chill}[\text{sp}]}$ ,  $\alpha_{\text{force}[\text{sp}]}$ ,  $\alpha_{\text{photo}[\text{sp}]}$ ) and individual species responses to cues ( $C_i$ ,  $F_i$ ,  $P_i$ , respectively). Finally, our model estimates the residual variation across species ( $\alpha_{\text{pheno}[\text{sp}]}$ ) and observations ( $\sigma_{\text{pheno}}$ ). Our model estimates the residual variation across species as well as the variation in cues not attributed to the trait (partial pooling).

$$\begin{aligned}\mu_{\text{pheno}} &= \alpha_{\text{pheno}[\text{sp}]} + \beta_{\text{chill}[\text{sp}]} \times C_i + \beta_{\text{force}[\text{sp}]} \times F_i + \beta_{\text{photo}[\text{sp}]} \times P_i \\ y_{\text{pheno}[i]} &\sim \text{normal}(\mu_{\text{pheno}}, \sigma_{\text{pheno}})\end{aligned}\quad (3)$$

$$\begin{aligned}\alpha_{\text{pheno}} &\sim \text{normal}(\mu_{\alpha_{\text{pheno}}}, \sigma_{\alpha_{\text{pheno}}}) \\ \alpha_{\text{force}} &\sim \text{normal}(\mu_{\alpha_{\text{force}}}, \sigma_{\alpha_{\text{force}}}) \\ \alpha_{\text{chill}} &\sim \text{normal}(\mu_{\alpha_{\text{chill}}}, \sigma_{\alpha_{\text{chill}}}) \\ \alpha_{\text{photo}} &\sim \text{normal}(\mu_{\alpha_{\text{photo}}}, \sigma_{\alpha_{\text{photo}}})\end{aligned}\quad (4)$$

Chilling was included as chill portions, and both forcing and photoperiod as continuous variables. We standardized all cues by *z*-scoring them, allowing us to make comparisons across cues (see Loughnan and Wolkovich (In prep) for more details).

For model, we used weakly informative priors unique to each trait. We validated our choice in priors using prior predictive checks. We coded our models in the Stan programming language and fit them using the rstan package. Each model had four chains, with models run for 6000-8000 total sampling

iterations. All our models met basic diagnostic checks, including no divergences, high effective sample sizes ( $n_{eff}$ ) that exceeded 10% of the number of iterations, and  $\hat{R}$  values close to 1. Our model estimates are reported as the mean values, with the 90% uncertainty interval. Estimates are given to one decimal place, as it best represents the level of precision we can infer from our datasets (Gelman et al., 2020).

## Results

Across our eight populations, we measured 47 species of which 28 were in our eastern transect and 22 within our western transect. These include species dominant in both the understory and canopy layer, with our eastern community consisting of 13 shrubs and 15 trees, and our western community consisting of 18 shrubs and 4 trees. There were three species that occurred in both transects. In total we measured traits of 1428 unique individuals between the two transects across our five traits: height ( $n = 1317$ ), DBH ( $n = 1220$ ), SSD ( $n = 1359$ ), LMA ( $n = 1345$ ), LNC ( $n = 1351$ ). Across our two growth chamber studies, we made observations of 4211 samples, with studies spanning 82 and 113 days for our eastern and western studies respectfully.

Most of our traits showed negligible variation across populations, both in terms of differences between the two transects (main effect of transect only) or by latitudes within each transect (an interactive effect between transect and latitude). Only leaf mass area (LMA) differed by latitude within transects (0.3, UI: 0.2,0.4)—with the LMA of eastern species increasing with latitude—but not western species (Fig. 1 d). Similarly, leaf nitrogen content (LNC) showed a weak interaction between latitude and transect (0, UI: 0.0,0)—with higher latitude populations in our eastern transect having lower LNC than lower latitude populations and western species (Fig. 1e). Our wood and structural traits, however, showed no differences across populations or transects (Table 2 - 3). The differences we found across populations were small to negligible, especially in comparison to species-level differences, which varied considerably and up to 33 fold (Fig. 2).

Only a subset of our traits related to at least one budburst cue, but responses were generally weak, with the exception of trait specific relationships to photoperiod. We found similar cue relationships for our two structural traits, with tall plants with larger DBH exhibiting stronger photoperiod responses and producing earlier estimates of budburst under longer daylengths (-1.7, UI: -2.9, -0.5 for height and -2.3, UI: -3.4, -1.1 for DBH). But we found no relationship between cues and SSD (Table 3) or between cues and species different types of wood porosity 3. Of our two leaf traits, only LMA showed a relationship to photoperiod, with low LMA species advancing their budburst timing in response to longer photoperiods (-7.5, UI: -10.9, -4.1). But we found no relationship between LNC and any of our three cues.

How our three cues shaped budburst timing also varied across our trait models. For most of our trait models, chilling followed by forcing, were the strongest cues of budburst (Fig. 4). But in accounting for the correlations between LMA and cues, our model estimated stronger responses to photoperiod (-7.5, UI: -10.9, -4.1) than forcing (-7.1, UI: -13.0, -1.6), and a relatively weak effect of chilling (90% UI crossing zero) (Table 4). In contrast, our LNC (Table 5) and SSD models (Table 3) both showed negligible relationships between photoperiod and budburst timing (90% UI crossing zero). These finding suggest that relationships with additional traits can alter the effects and relative importance of cues on budburst (Fig. 5). This—paired with strong species-level variation—may cause variation in the estimated responses of species relative budburst dates across the different trait models.

In synthesizing the effects of multiple traits across species, our results can be used to make generalizations across ecologically important groups of species. But only some of our models estimated clear

gradients in species timing between trees and shrubs (Fig. 6). In particular, we found height to have strong correlations between budburst timing and trait values, with earlier estimates of budburst for shrubs—especially under stronger cues—and later budburst estimates for trees (Fig. 6). Diameter at breast height showed similar trends as estimates from the height model (results not shown). But this was not the case for our two leaf traits. Leaf nitrogen content, for example, showed no distinct separation between shrub and tree functional groups (Fig. 6).

## Discussion

### 1. General summary of what we FOUND (with some of what we did)

- (a) Our study = one of the first to combine trait data with phenological cue responses for the same individuals and across species distributions
- (b) We found only three (out of six) traits showed a relationship with phenological traits, with two of those fitting our hypothesized aqius to cons paradigm in the spring.
- (c) Joint modelling approach = use sp-level estimates for traits to understand phenological cue responses and budburst timing suggested photoperiod may be most important to trait-phenology relationships ..
- (d) Includes plant communities in eastern and western deciduous forests of North America, and samples collected from multiple populations—modelling approach that accounts for variation across populations and transects
- (e) We found, however, generally low population-level variation in traits—usually even across the different eastern and western deciduous forests of North America, which was likely driven by species-level variation in traits

### 2. Population level variation:

- (a) generally traits did not change across population or with latitude
- (b) Only leaf traits varied with transect—patterns are in line with what you would predict given the differences in shrub vs tree spp in our transects
- (c) Predicted shrubs = acquisitive growth strategies = low LMA and high LNC
- (d) Western pop = more shrubs = did have low LMA and high LNC relative to eastern transect

### 3. Phenology in context of LES

- (a) Summarize whether fits with acquisitive-conservative predictions
  - i. Partial support whether phenology = the same acquisitive to conservative gradient in growth commonly found for other traits
  - ii. Predicted all traits to correlate with phenological cues = budburst timing—but found relationships with only one of our three cues
  - iii. Ht and DBH both in line without prediction that tall, larger trees would have stronger photoperiod cues
  - iv. But LMA showed opposite response to photoperiod than predicted—low LMA (acquisitive) spp more responsive to photoperiod
- (b) Interesting similarities with previous studies
  - i. Traits have many functions—mitigate abiotic and biotic factors
  - ii. But trait research has found considerable variation in trait phenotypes across spatial scales (cite Gross 2000, Flecker and Jones 2003)—different responses for different traits

- iii. In comparing plant traits across spatial scales, previous studies have found little variation across regional scales (comparing populations or sites) and more within-community variability (Messier et al 2010, Kang et al. 2014, Standen et al. 2023)
- iv. But in comparing our results with a global meta-analysis of trait relationships with budburst cues = similar results for some traits
- v. Found similar unexpected responses in leaf traits—high specific leaf area (the inverse of LMA) exhibited strong responses to photoperiod—suggests could be alternate underlying mechanisms shaping how species respond to photoperiod cues
- vi. But also found strong temperature cues for most traits—not found in our analysis of North American forests

## How traits shape species temporal niches

### 4. Correlations between traits and cues

- (a) Individuals temporal niche = result of numerous and likely complex interactions across abiotic and biotic factors
- (b) Our results suggest focusing on single traits—fails to account for interactions between other traits and cues
- (c) Decades of research on budburst cues = primarily driven by temperature cues—but only photoperiod related to other traits

### 5. Suggests may be other physiological constraints or trade-offs selecting for other traits and shaping species phenologies

- (a) Temperature—generally thought of as a strong environmental cue—but might not be strongest selective pressure in temperate forests
- (b) Soil moisture = known to select for other types of tree phenology—radial growth and shoot elongation
- (c) LMA can also vary with soil moisture—to reduce evaporation plants can reduce leaf area = high LMA in dry conditions (de la Riva, 2015)—could be driving the unexpected trend in LMA
- (d) Later in the spring—longer photoperiods—less water high LMA would be favorable
- (e) To gain a holistic understanding of species growth strategies and phenotypes require consideration of additional cues

## Using functional traits to predict climate change responses

### 6. Impacts of community dynamics and forecasting

- (a) Community dynamics with climate change
  - i. Results = novel insights on how other traits impact phenological sensitivities to environmental cues
  - ii. Relationships between budburst and photoperiod could limit the extent of spp advances in spring
  - iii. Photoperiod is fixed—unlike temperature cues = increasing with climate change—species ability to track temperature could be constrained
- (b) Forecasting future forests

- 243 i. Our study is one of the first to model the relationship between phenology and other  
244 traits—but results simply tell us the associations—need to understand mechanism
- 245 ii. Phenology—cited as finger print of climate change—but other traits have been noted  
246 to change as well
- 247 iii. Across temperature and precipitation gradients—leaf size and shape also change—spp  
248 shift to conserve water and mitigate effects of transpiration with high high temperatures—  
249 cite de le Riva 2016 as example
- 250 iv. trait literature = explored how traits shift across environmental gradients—temp or  
251 precip—denoting trends leaf area and height
- 252 v. Important we continue to synthesize across phenology and trait research—better fore-  
253 cast the future impacts of climate change on communities

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## Figures

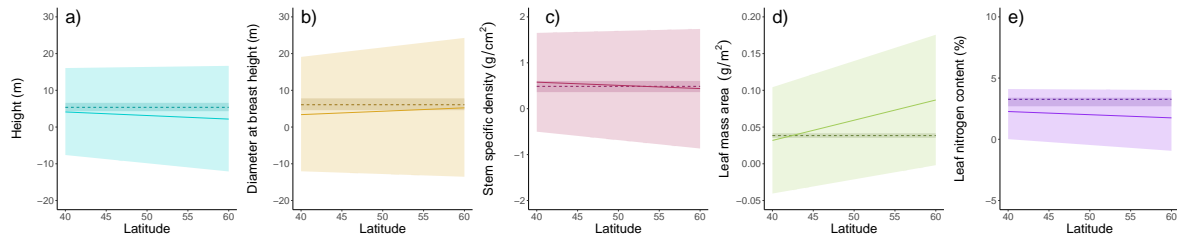


Figure 1: We found geographic differences for four of our functional traits, with the direction of the relationship varying across traits. Of our traits a. height, b. diameter at breast height, c. leaf mass area, and e. leaf nitrogen content showed strong interactions between latitude and transect, while d. stem specific density showed no effects of geography. Dashed lines represent the western transect and solid lines the eastern transect.

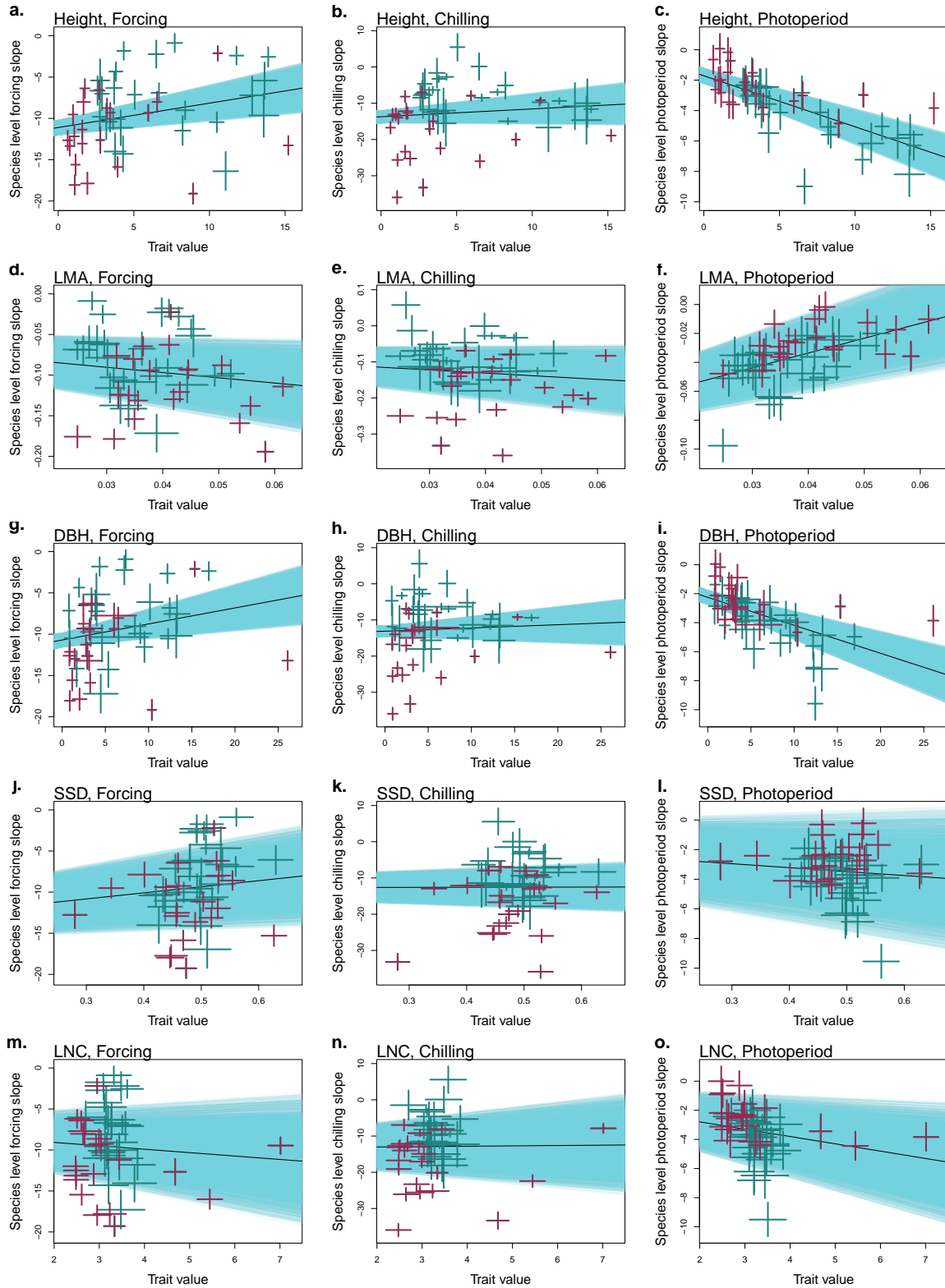


Figure 2: Relationship between species traits and cue responses, for height (a-c), leaf mass area (d-f), diameter at breast height (g-i), stem specific density (j-l), and the leaf nitrogen content (m-o). Point colours representing different species groups, with tree species are depicted in maroon and shrub species in teal.

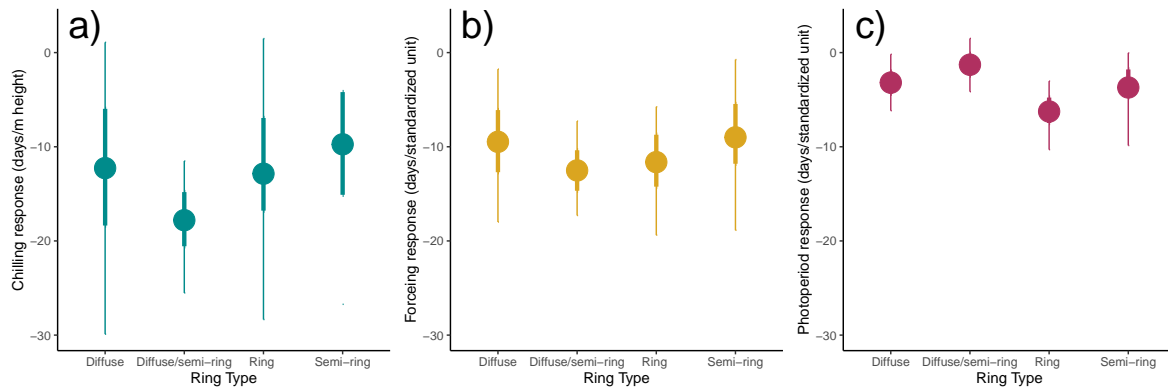


Figure 3: Despite species wood structures causing differing growth strategies, we did not find this trait to correlate with differences in cue responses across species. Thinner lines represent the 90% UI and thicker lines the 50% UI. Here we show the results for height only.

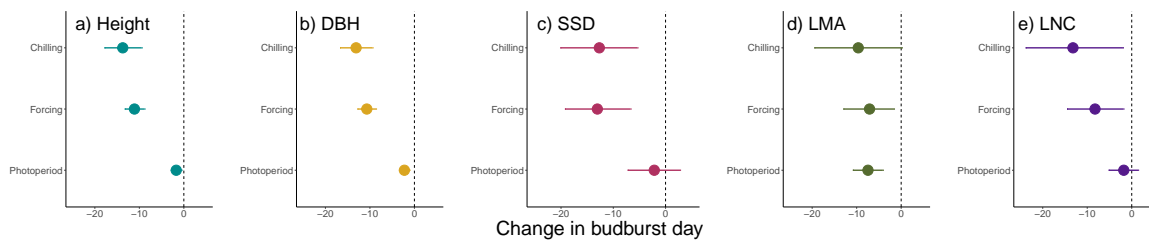


Figure 4: We found fairly consistent estimates for budburst cue responses to chilling, forcing, and photoperiod for each of our trait models: a. height, b. diameter at breast height, c. stem specific density, d. leaf mass area, and e. leaf nitrogen content. Lines represent 90% uncertainty intervals.



Figure 5: The relationships between traits and cue responses varied considerably across each of our trait models, a. height, b. diameter at breast height, c. stem specific density, d. leaf mass area, and e. leaf nitrogen content, and for individual cues. Lines represent 90% uncertainty intervals. Note the differences in the scale of the x-axis.

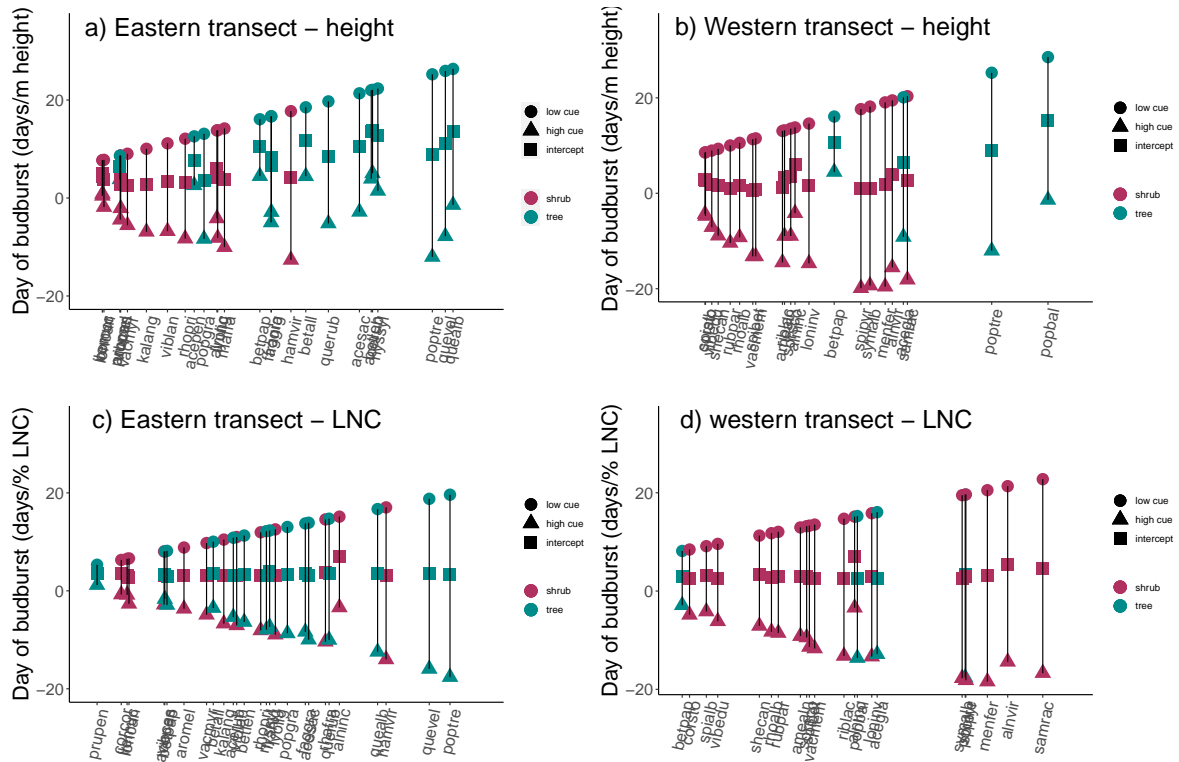


Figure 6: We found budburst estimates differed between our full model (intercept plus cues, depicted as triangles for high cues and as circles for low cues), versus the intercepts only model (without cues, shown as squares). Species are ordered in increasing budburst dates for both the eastern (a and c) and western (b and d) populations, spanning from early budbursting shrubs, in red, to late budbursting trees in blue. For traits such as height (a and b) we found distinct partitioning of budburst across shrub and tree species, but this was not the case for all traits, with our model of leaf nitrogen content showing highly mixed budburst order of shrub and tree species (c and d).

## Tables

Table 1: Summary output from a joint Bayesian model of height and budburst phenology in which species are partially pooled. The effect of transect is modeled as a categorical variable and latitude as continuous in interaction term with transect. The model includes photoperiod and forcing as continuous variables, with all three cues  $z$ -scored to allow comparisons across cues.

	mean	5%	25%	75%	95%
Transect	2.60	-3.00	0.40	4.90	8.30
Transect x latitude	-0.10	-0.20	-0.10	-0.00	0.00
Forcing	-11.10	-13.30	-12.00	-10.20	-8.80
Chilling	-13.70	-17.80	-15.50	-12.10	-9.50
Photoperiod	-1.70	-2.90	-2.20	-1.20	-0.50
Trait x forcing	0.30	-0.10	0.20	0.40	0.60
Trait x chilling	0.20	-0.50	-0.00	0.50	0.90
Trait x photoperiod	-0.30	-0.50	-0.40	-0.30	-0.20

Table 2: Summary output from a joint Bayesian model of DBH and budburst phenology in which species are partially pooled. The effect of transect is modeled as a categorical variable and latitude as continuous in interaction term with transect. The model includes photoperiod and forcing as continuous variables, with all three cues  $z$ -scored to allow comparisons across cues.

	mean	5%	25%	75%	95%
Transect	-6.40	-13.70	-9.40	-3.40	1.00
Transect x latitude	0.10	-0.10	0.00	0.20	0.30
Forcing	-10.70	-12.80	-11.60	-9.90	-8.60
Chilling	-13.10	-16.70	-14.60	-11.60	-9.40
Photoperiod	-2.30	-3.40	-2.70	-1.80	-1.10
Trait x forcing	0.20	-0.10	0.10	0.30	0.40
Trait x chilling	0.10	-0.40	-0.10	0.30	0.50
Trait x photoperiod	-0.20	-0.30	-0.20	-0.10	-0.10

Table 3: Summary output from a joint Bayesian model of SSD and budburst phenology in which species are partially pooled. The effect of transect is modeled as a categorical variable and latitude as continuous in interaction term with transect. The model includes photoperiod and forcing as continuous variables, with all three cues  $z$ -scored to allow comparisons across cues.

	mean	5%	25%	75%	95%
Transect	0.40	-0.10	0.20	0.60	0.90
Transect x latitude	-0.00	-0.00	-0.00	-0.00	0.00
Forcing	-13.00	-19.30	-15.70	-10.50	-6.70
Chilling	-12.70	-20.10	-15.60	-9.70	-5.40
Photoperiod	-2.20	-7.20	-4.20	-0.00	2.80
Trait x forcing	7.40	-5.70	2.00	13.00	20.10
Trait x chilling	0.30	-15.60	-5.90	6.50	15.90
Trait x photoperiod	-2.70	-12.60	-7.00	1.50	7.80

Table 4: Summary output from a joint Bayesian model of LMA and budburst phenology in which species are partially pooled. The effect of transect is modeled as a categorical variable and latitude as continuous in interaction term with transect. The model includes photoperiod and forcing as continuous variables, with all three cues  $z$ -scored to allow comparisons across cues.

	mean	5%	25%	75%	95%
Transect	-11.70	-15.30	-13.20	-10.20	-8.00
Transect x latitude	0.30	0.20	0.20	0.30	0.40
Forcing	-7.10	-13.00	-9.30	-4.80	-1.60
Chilling	-9.60	-19.50	-13.50	-5.70	0.20
Photoperiod	-7.50	-10.90	-8.90	-6.00	-4.10
Trait x forcing	-0.60	-2.00	-1.20	-0.10	0.80
Trait x chilling	-0.90	-3.40	-1.90	0.10	1.70
Trait x photoperiod	1.00	0.20	0.70	1.40	1.90

Table 5: Summary output from a joint Bayesian model of LNC and budburst phenology in which species are partially pooled. The effect of transect is modeled as a categorical variable and latitude as continuous in interaction term with transect. The model includes photoperiod and forcing as continuous variables, with all three cues  $z$ -scored to allow comparisons across cues.

	mean	5%	25%	75%	95%
Transect	0.00	-0.80	-0.30	0.40	0.90
Transect x latitude	-0.00	-0.00	-0.00	-0.00	-0.00
Forcing	-8.20	-14.50	-10.80	-5.80	-1.80
Chilling	-13.20	-23.80	-17.70	-8.80	-2.00
Photoperiod	-1.80	-5.20	-3.10	-0.40	1.40
Trait x forcing	-0.40	-2.40	-1.20	0.40	1.40
Trait x chilling	0.10	-3.30	-1.20	1.50	3.20
Trait x photoperiod	-0.50	-1.40	-0.90	-0.10	0.50