

Understanding growing degree days to predict spring phenology in a warming world

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Introduction

1. Need to start even bigger here! In ecology, we have the fundamental issue of understanding and applying methods to accurately predict shifts in climate and the broader impacts of these shifts.
 - (a) Often we use mixed models to answer ecological questions, though we do not always understand the intricacies of the model output, nor do we investigate what is missing from the model output.
 - (b) Here, we work to understand mixed models using simulation data and test myriad hypotheses through these simulations.
 - (c) These methods can be applied to many ecological questions investigating climate data across global habitats but here we will investigate the effects of climate measurements and site on spring plant phenology.
2. Understanding and predicting plant phenology in temperate deciduous forests is critical as it both shapes community structure and also influences major ecosystem services such as resource and forest management.
 - (a) Climate change and urbanization are advancing spring timing—such as budburst and leafout, which are strongly cued by temperature, resulting in longer growing seasons (Chuine *et al.*, 2001) which ultimately impacts these services.
 - (b) Temperate forests sequester carbon and help mitigate the negative effects of climate change and—with earlier spring phenology and longer growing seasons—there has been an increase in carbon uptake (Keenan *et al.*, 2014).

- (c) But our understanding of how climate change is impacting this timing of spring is incomplete, especially in urban versus natural forest habitats.
3. Urbanization has led to the formation of urban heat islands, which have been shown to affect plant phenology and lead to earlier spring leafout (Meng *et al.*, 2020).
- (a) These trends are crucial to understand in order to predict plant development with warming.
 - (b) Tracking heat accumulation is one way to measure and forecast spring leafout, which is often predicted through the growing degree day (GDD) model (Cook *et al.*, 2012; ?; Phillimore *et al.*, 2013; Schwartz *et al.*, 2006; Vitasse *et al.*, 2011).
 - (c) The GDD model simply sums temperatures above a certain threshold—ideally around 0°C as estimates are proven to be more accurate (Man & Lu, 2010)—and different species often require a different number of GDDs to leaf out.
 - (d) GDDs accumulate at a faster rate when mean temperatures are higher, thus different sites or different climate measurement methods may record different GDD thresholds for leafout.
 - (e) Spring leafout timing can have cascading effects to pollinators (Boggs & Inouye, 2012; Pardee *et al.*, 2017), on carbon dynamics (Richardson *et al.*, 2013) and albedo (Williamson *et al.*, 2016), thus integrating the growing degree day model successfully is essential for predicting the effects of climate change on temperate systems.
4. Phenology is often measured through satellite, remote sensing or PhenoCam images to detect spring ‘green-up’ (Meng *et al.*, 2020; ?; Richardson, 2015) but these methods fail to detect the species—or even site-level—nuances in leafout timing (Elmendorf *et al.*, 2019).
- (a) Intensive, on the ground observations of individual budburst and leafout timing is the most effective way to implement new methods in calculating growing degree days and predicting future phenology.
 - (b) Urban environments additionally provide a natural laboratory for assessing the effects of warming on temperate tree and shrub species as these sites are warming at a faster rate than more rural habitats (Pickett *et al.*, 2011; Grimm *et al.*, 2008).
5. Now I want to talk about how arboreta offer another unique lens by incorporating varying provenances and seed sources.
- (a) I want to set up this hypothesis here
6. Here I will talk about the differences between using hobo loggers and weather stations
- (a) I want to set up the two hypotheses here about temperature accuracy.

7. Here, we use both simulations, models and real data to test our hypotheses on modelling GDD accuracy in a warming world.
 - (a) Urban environments require fewer GDDs to leafout than forest habitats.
 - (b) Individuals with provenance latitudes from more northern locations require fewer GDDs to leafout.
 - (c) Hobo loggers are less accurate measures of the same weather as weather stations.
 - (d) Hobo loggers better capture urban or provenance effects.

Methods

Sites

1. We chose two sites—one urban arboretum and one forest—with similar species and climates to compare the number of growing degree days to leafout across species.
 - (a) The urban site is in Boston, MA at the Arnold Arboretum of Harvard University ($42^{\circ}17'$ N $-71^{\circ}8'$ W).
 - (b) The Arnold Arboretum is 281 acres and has a range of elevation of XX to XX
 - (c) The forest site is in Petersham, MA at the Harvard Forest ($42^{\circ}31'53.5''$ N $-72^{\circ}11'24.1''$ W).
 - (d) The Harvard Forest is XXX acres and has a range of elevation of XX to XX.

Simulations

Shiny App

Data analysis

1. Using Bayesian hierarchical models with the rstan package (Stan Development Team, 2019), version 2.19.2, in R (R Development Core Team, 2017), version 3.3.1, we estimated the effects of urban or provenance effect and method effect and all two-way interactions as predictors on GDDs until leafout.
 - (a) Species were modeled hierarchically as grouping factors, which generates an estimate and posterior distribution of the overall response across the XX species used in our simulations.
 - (b) We ran four chains, each with 2 500 warm-up iterations and 4 000 sampling iterations for a total of 6 000 posterior samples for each predictor for each model using weakly informative priors.

- (c) Increasing priors three-fold did not impact our results.
- (d) We evaluated our model performance based on \hat{R} values that were close to one and did not include models with divergent transitions in our results.
- (e) We also evaluated high n_{eff} (4000 for most parameters, but as low as 1400 for a couple of parameters in the shoot apical meristem model).
- (f) We additionally assessed chain convergence and posterior predictive checks visually (Gelman *et al.*, 2014).

Real Data

1. Phenology observations across the Arnold Arboretum were collected by trained citizen scientists from the Tree Spotters National Phenology Network program (?).
 - (a) The Tree Spotter volunteers observed 15 species with varying phenologies and each species had 5 individuals for a total of 75 trees.

Results

- i. Urban environments require fewer GDDs to leafout than forest habitats.
- ii. Individuals with provenance latitudes from more northern locations require fewer GDDs to leafout.
- iii. Hobo loggers are less accurate measures of the same weather as weather stations.
- iv. Hobo loggers better capture urban or provenance effects.
- v. Shiny App
- vi. Real data

Discussion

- i. Add section here that discusses why maybe using GDD models may not be appropriate for the future with warming (Man & Lu, 2010).
 - A. This is because with warming, GDDs will accumulate at a faster rate, which will reduce accuracy of determining that actual threshold for leafout phenology.
 - B. In the future, we need to either use a method that is less reliant on accumulated sums—especially if it is a climatological sum—or we must scrutinize results through the use of mixed models and simulated data as we demonstrate here.

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Tables and Figures