Soil moisture interacts with temperature to affect plant phenology

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Author contributions: All authors conceived of this manuscript, which began at a Radcliffe Exploratory Seminar in 2016, and all authors contributed to manuscript revisions. AKE and EMW conceived of the idea for the literature review, database compilation, and related Radcliffe Exploratory Seminar. AKE compiled the datasets; AKE analyzed the data and created the figures.

Data Accessibility The data reported in this paper are from the MC3E and ExPhen databases, which are

available at KNB (Ettinger and Wolkovich, 2018; ?)

Running title Soil moisture affects phenology

**Key words** global warming, warming experiment, microclimate, phenology, bud-burst, leaf-out, flowering, fruiting, senescence

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#### 1 Abstract

Past studies of phenology responses to recent climate change have focused on temperature as a driver of observed shifts. However, soil moisture is also affected by climate change and likely to alter biological responses. Here we synthesize microclimate and phenology data from climate change experiments to show how soil moisture interacts with temperature to affect plant phenology. We find that soil drying generally delays plant phenology, especially for budburst, a rate of 0.42 days per percent reduction in soil VWC, on average. These effects vary widely across sites and species. This rate suggests that, on average, climate change-induced shifts soil moisture will play only a small role in altering future phenology, compared to shifts in temperature. This is both because of the strong sensitivity of plant phenology to temperature and because of the large magnitude of projected shifts in temperature, compared to shifts in soil moisture.

Nonetheless, although effects of soil moisture are comparably small on average, sensitivity to soil moisture varied dramatically by species, and soil moisture levels differed by site and among years. Thus, soil moisture is likely to play a role in phenological shifts with climate change for some species, in some locations and years. Quantifying phenological sensitivity to changes in soil moisture will be useful for improving forecasts of shifts in phenology with future climate change.

#### $_{ ext{ iny 16}}$ $\operatorname{INTRODUCTION}$

Climate change is affecting organisms by altering temperature and soil moisture around the world (Parmesan, 2006; ?). One of most widespread biological responses to climate change is a shift in phenology, the timing of recurring biological events, which has occurred at rates of 2.3-5.1 days per decade (Parmesan, 2006; 19 Poloczanska et al., 2013; Root et al., 2003). Shifts in plant phenology are the most widely documented, with spring phenology (budburst, leaf-out, and flowering) occurring earlier in recent years (Wolkovich et al., 2013), 21 and senescence occurring later (Taylor et al., 2008; Delpierre et al., 2009). Phenological shifts are typically attributed to warming temperature, a known and well-studied driver of plant 23 phenology. The timing of spring budburst, for example, depends on temperature through both chilling (the prolonged exposure to cold temperatures after growth cessation in the fall) and forcing (exposure to warm temperatures). Recent trends of advancing phenology may be due to increases in both/either chilling and/or forcing with global warming (Fujisawa and Kobayashi, 2010; Ibanez et al., 2010; Cook et al., 2012). In places where delays in spring phenology have occurred, reductions in winter chilling are thought to be the cause (Yu et al., 2010). Effects of altered precipitation and soil moisture on phenology have received less attention, but are likely to 30 be important drivers of plant phenology. For example, budburst, flowering, and leaf drop are affected by tree 31 water status in dry ecosystems (Essiamah and Eschrich, 1986; Reich and Borchert, 1984; van Schaik et al., 32 1993). Budburst can be slowed by water stress through inhibiting cell elongation (Essiamah and Eschrich, 1986), and growing season start may be delayed by drought in grasslands Cui et al. (2017). Flowering phenology, on the other hand, can be advanced by drought conditions (Hamann et al., 2018). Here we conduct a meta-analyses of climate change experiments to test whether and how soil moisture interacts with temperature to affect plant phenology. Field-based climate change experiments that warm plots to different levels offer valuable tools to study climate change impacts on plant phenology. Experiments 38 can combine temperature and precipitation treatments to create the "no-analog" climate scenarios forecasted 39 for the future, particularly when they employ active-warming methods, such as forced air heaters, soil warming cables, or infrared heaters (Shaver et al., 2000; Williams et al., 2007; Aronson and McNulty, 2009). Climate

change experiments often monitor daily soil moisture, as well as daily air and soil temperature, at the plot-

- 43 level, allowing detailed quantification of how microclimate affects plant phenology.
- Previous meta-analyses of phenology in climate change experiments have focused primarily on effects of temperature (Wolkovich et al., 2013). We expected that soil moisture may also affect phenology, with
- drier soils delaying budburst and leafout phenology and advancing flowering and fruiting phenology. We use
- 47 measured microclimate and phenology data from two databases of climate change experiments: MicroClimate
- 48 from Climate Change Experiments (MC3E) and Experimental Phenology (ExPhen) to quantify effects of
- 49 soil moisture and above-ground temperature on plant phenology (bud-burst, leaf-out, flowering, fruiting,
- 50 senescence; see Materials and Methods). We also use forecasted changes in temperature and soil moisture to
- investigate how including soil moisture alters expected future shifts in phenology.

#### 52 MATERIALS AND METHODS

- 53 Data- To investigate how soil moisture interacts with temperature to affect phenology, we used two databases
- that compile data from climate change experiments. Microclimate data came from the MicroClimate from
- <sup>55</sup> Climate Change Experiments (MC3E) database Ettinger and Wolkovich (2018). Phenology data came from
- <sup>56</sup> a ExPhen, a new database of phenology from climate change experiments ?.
- 57 To create both databases, we first identified published, active-warming field experiments, many of which
- included precipitation manipulations. We focused on in situ active-warming manipulations because recent
- analyses indicate that active-warming methods are the most controlled and consistent methods available
- 60 for experimental warming (Kimball, 2005; Kimball et al., 2008; Aronson and McNulty, 2009; Wolkovich
- 61 et al., 2013). We carried out a full literature review to identify potential active-warming field experiments,
- 62 following the methods and search terms of Wolkovich et al. (2013) for their Synthesis of Timings Observed in
- iNcrease Experiments (STONE) database (Wolkovich et al., 2013), but restricting our focus to active-warming
- experiments. Further, because our goal was to tease out variation in microclimate (including temperature and
- soil moisture), we focused on warming studies that included both/either multiple levels of warming and/or
- 66 precipitation treatments. These additional restrictions constrained the list to 11 new studies published after
- 67 the STONE database, as well as six of the 37 studies in the STONE database. We contacted authors to
- obtain daily microclimate and phenological data for these 17 studies and received data (or obtained publicly

available data) for 10 of them, as well as datasets from five additional sites offered or suggested to us over the
course of our literature review and data analysis. The daily temperature and soil moisture data from these 15
experiments comprise the MC3E database, which is available at KNB (Ettinger and Wolkovich, 2018). The
phenology data from the same 15 experiments comprise the ExPhen database of experimental phenology,
which is also available at KNB (?).

Analysis—To understand how soil moisture interacts with temperature to affect phenology, we fit models with
measured soil moisture, measured temperature, and their interaction to phenology response data (budburst,
leafout, flowering, fruiting, senescence). Microclimate data came from the MC3E database, and phenology
data came from the ExPhen database. We excluded conifers from the analysis, because their phenology has
distinct differences from angiosperm phenology Polgar et al. (2014) and conifer data existed from only one site
in the database. For all phenology variables, the response variable was day-of-year of the phenological event.

Predictors for our primary models were measured air temperature, soil moisture, and their interaction. We
also fit model(s) with soil temperature in place of air temperature. Random effects for all phenology models
were species (with random slopes and intercepts), site (random intercept), and year nested within site (random
intercept). Equations for these models can be found in the Supplemental Methods. To better understand
the interactive effects of measured temperature and soil moisture, we conducted follow-up analyses in which
we fit the same phenology models subsets of the data (controls only, low temperature treatments only, and
high temperature treatments only).

cut This, Right? To quantify how climate manipulations affect temperature and soil moisture, we used microclimate data from the 4 sites in the MC3E database that manipulated both precipitation and temperature, and measured both above-ground temperature and soil moisture data (exp01,exp05,exp07,exp12). (OR use all 8 studies that measured soil moisture and above-ground temp?) We then fit two groups of hierarchical models to microclimate data from these sites: one group with temperature response variables (including mean annual, and mean seasonal temperatures), and the other group with soil moisture as response variables (including mean annual, and mean seasonal soil moisture). For both groups of models, explanatory variables were temperature treatment, precipitation treatment, and their interaction. The models included a random effect of site, with a random slope and intercept structure, to allow effects of experimental treatments to vary across sites. Models also included random effects of day-of-year, nested within year, on the intercept only, to

- 97 account for non-independence of measurements taken on the same day within the same year. Equations for
- these models can be found in the Supplemental Methods.

## 99 RESULTS

- 100 How do soil moisture and temperature affect phenology? We find that soil drying delays phenology
- and warming temperatures advance phenology, for most phenophases. The magnitude of these effects varies
- 102 across phenophases, species, and sites.
- Effects of soil moisture were strongest for budburst and leafout, and affected all phenophases except fruiting
- <sup>104</sup> (Figures 1, 1S). Soil drying delays spring budburst at a rate of 0.42 days per percent reduction in soil VWC.
- Thus, if soil moisture is reduced by 10% of its current state (mean across all sites= XX), as is expected over
- the next 50 years in the northeastern US (Berg et al., 2017) budburst would be delayed by approximately
- 107 XX days on average, due to changes in soil moisture alone.
- 108 Increasing air temperature advanced phenology for all phenophases except senescence (Figure 1S). Warming
- advances budburst phenology at a rate of 3.42 days per °C, a rate that is consistent with estimates from
- previous meta-analyses (Wolkovich et al., 2013).

## 111 DISCUSSION

- Soil moisture affects phenology, budburst, and leaf-out (not previously a focus of experiments and meta-
- analyses).
- 114 Variable responses to moisture (and precip) may be caused by temporal and spatial variation in the most
- limiting resource (temperature vs moisture), as well as differences across species
- As global warming reduces temperature limitation, importance of moisture limitation in plant phenology may
- increase.

# 118 Conclusions

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# Figures

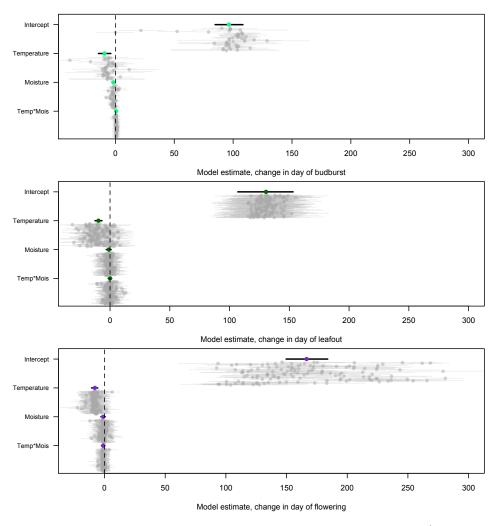


Figure 1: Model coefficients from budburst, leafout, and flowering models (with centered predictors).