

¹ Regional Risk Outline

² Authors:

³ C. J. Chamberlain ^{1,2}, B. I. Cook ³, I. Morales Castilla ^{1,4} & E. M. Wolkovich ^{1,2}

⁴ Author affiliations:

⁵ ¹Arnold Arboretum of Harvard University, 1300 Centre Street, Boston, Massachusetts, USA;

⁶ ²Organismic & Evolutionary Biology, Harvard University, 26 Oxford Street, Cambridge, Massachusetts, USA;

⁷ ³NASA Goddard Institute for Space Studies, New York, New York, USA;

⁸ ⁴Edificio Ciencias, Campus Universitario 28805 AlcalÁ de Henares, Madrid, Spain

⁹ *Corresponding author: 248.953.0189; cchamberlain@g.harvard.edu

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¹¹ Introduction

¹² 1. Temperate tree and shrub species are at risk of damage from late spring freezing events, also known as
¹³ false springs, and this risk may shift with climate change

¹⁴ (a) The growing season is lengthening (mainly due to earlier springs) across many regions in the
¹⁵ northern hemisphere (Chen *et al.*, 2005; Liu *et al.*, 2006; Kukal & Irmak, 2018), but last spring
¹⁶ frosts still pose a threat in many of these regions (Wypych *et al.*, 2016).

¹⁷ (b) Spring onset is advancing with temperate tree and shrub species initiating leafout 4-6 days on
¹⁸ average earlier per °C of warming (Wolkovich *et al.*, 2012; IPCC, 2015).

¹⁹ (c) Last spring freeze dates are not predicted to advance at the same rate (Inouye, 2008; Martin *et al.*,
²⁰ 2010; Labe *et al.*, 2016; Sgubin *et al.*, 2018), potentially amplifying the effects of false spring events
²¹ in these regions.

²² (d) In Germany, for example, the last freeze date has advanced by 2.6 days per decade since 1955
²³ (Zohner *et al.*, 2016) but leafout is advancing around twice as fast.

²⁴ (e) Major false spring events have been recorded in recent years and have found it can take 16-38
²⁵ days for trees to refoliate (Gu *et al.*, 2008; Augspurger, 2009, 2013; Menzel *et al.*, 2015), which can
²⁶ detrimentally affect crucial processes such as carbon uptake and nutrient cycling (Hufkens *et al.*,
²⁷ 2012; Richardson *et al.*, 2013; Klosterman *et al.*, 2018).

²⁸ 2. Episodic frosts are one of the largest limiting factors in species range limits and have shaped plant life
²⁹ history strategies (Kollas *et al.*, 2014).

- 30 (a) Temperate plants are exposed to freezing temperatures numerous times throughout the year,
31 however, individuals are most at risk to damage from stochastic spring frosts, when frost tolerance
32 is lowest (Sakai & Larcher, 1987).
- 33 (b) Frost tolerance greatly diminishes once individuals exit the dormancy phase (i.e. processes leading
34 to budburst) through full leaf expansion (Vitasse *et al.*, 2014; Lenz *et al.*, 2016).
- 35 (c) Individuals that initiate budburst and have not fully leafed out before the last spring freeze are
36 at risk of leaf tissue loss, damage to the xylem, and slowed canopy development (Gu *et al.*, 2008;
37 Hufkens *et al.*, 2012).
- 38 (d) Individuals and species that initiate budburst earlier in the season are more frost resistant (Körner
39 *et al.*, 2016), however, as climate change advances, less frost resistant individuals may start initiating
40 budburst before the last freeze date.
- 41 (e) Despite the importance of false spring events, the extent of damage and the frequency and intensity
42 of false spring events is still largely unknown.
- 43 3. Plant phenology and climate are tightly intertwined (Pau *et al.*, 2011), thus understanding major
44 climate oscillation patterns is essential for predicting budburst and, ultimately, false spring risk.
- 45 (a) The North Atlantic Oscillation (NAO) index is often used to describe winter and spring circulation
46 across Europe.
- 47 (b) More positive NAO phases tend to result in higher than average winter and spring temperatures,
48 and with climate change, higher NAO phases has correlated to even earlier budburst dates in
49 some regions (Chmielewski & Rötzer, 2001), however it is unclear if more positive NAO phases
50 also translates to more false springs.
- 51 4. There is large debate over whether or not spring freeze damage will increase (Hänninen, 1991; Augspurger,
52 2013; Labe *et al.*, 2016), remain the same (Scheifinger *et al.*, 2003) or even decrease (Kramer, 1994;
53 Vitra *et al.*, 2017) with climate change and there is also great variation within studies.
- 54 (a) Some research suggests false spring incidence has declined in many regions (i.e. across parts of
55 North America and Asia), however the prevalence of spring frosts has consistently increased across
56 Europe since 1982 (Liu *et al.*, 2018).
- 57 (b) However, recent studies have demonstrated regional effects may be more closely related to false
58 spring risk: whether via altitudinal variation (Vitra *et al.*, 2017; Ma *et al.*) or distance from the
59 coast (Wypych *et al.*, 2016; Ma *et al.*).
- 60 (c) By better understanding these regional climatic implications and which factors are most crucial
61 for predicting risk, we may be able to determine which regions may be at risk currently and which
62 regions may become more at risk in the future.

- 63 5. The majority of false spring studies assess the effects of one predictor (e.g. temperature, elevation or
64 distance from the coast) on false spring prevalence but most fail to incorporate multiple effects.
- 65 (a) Our primary aim is to investigate the known regional factors on false spring risk and compare the
66 effect of them and their interaction with climate change. The key regional factors we identify for
67 this study are: mean spring temperature, NAO index, elevation and distance from the coast.
- 68 6. By refining and identifying budburst and climate trends in recent years, we could improve future
69 projections in false springs.
- 70 (a) For this purpose, we assessed the number of false springs that occurred across 11,648 sites around
71 Europe, spanning altitudinal and coastal gradients, using observed phenological data (754,786
72 observations) for six temperate, deciduous trees and combined that with daily gridded climate
73 data for each site that extended from 1951-2016.
- 74 (b) In this study, a false spring was tallied when temperatures fell below -2.2° (Schwartz, 1993) between
75 budburst and leafout (CITE Rethinking here?).
- 76 (c) Since the primary aim of the study is to predict false spring incidence in a changing climate, we
77 split our data to before and after 1983 to capture reported temporal shifts in temperature trends
78 (Stocker *et al.*, 2013; Kharouba *et al.*, 2018).
- 79 (d) We predicted that: (1) Earlier budburst species would experience more false springs, especially
80 after 1983 and (2) there would be different regional effects (i.e. mean spring temperature, NAO
81 index, elevation, distance from the coast) on false spring incidence and those trends would shift
82 when coupled with the effects of climate change.

83 Methods

84 Phenological Data and Calculating Vegetative Risk

- 85 1. We obtained phenological data from the Pan European Phenology network (PEP725, www.pep725.edu),
86 which provides open access phenology records across Europe (Templ *et al.*, 2018).
- 87 2. Since plants are most susceptible to damage from frost between budburst and full leafout, we selected
88 only leafout data (i.e., in Meier, 2001, BBCH 11, which is defined as the point of leaf unfolding and the
89 first visible leaf stalk) from the PEP725 dataset.
- 90 3. We then subtracted 12 days from the leafout date to establish a rough estimate for day of budburst
91 (Donnelly *et al.*, 2017).

- 92 4. The species used in the study were *Aesculus hippocastanum* Poir., *Alnus glutinosa* (L.) Gaertn., *Betula*
93 *pendula* Roth., *Fagus sylvatica* Ehrh., *Fraxinus excelsior* L., *Quercus robur* L.
- 94 5. Selection criteria for the species were as follows: (1) to be temperate, deciduous species that were not
95 cultivars or used for crops, (2) there were at least 90,000 observations of BBCH 11, (3) to represent
96 over half of the total number of sites available (11,684), (4) there were observations for at least 65 out
97 of the 66 years of the study (1951-2016).

98 **Climate Data**

- 99 1. We collected daily gridded climate data from the European Climate Assessment & Dataset (ECA&D)
100 and used the E-OBS 0.25 degree regular latitude-longitude grid from version 16.
- 101 2. We used the daily minimum temperature dataset to determine if a false spring occurred.
- 102 3. False springs in this study were defined as temperatures at or below -2.2°C (Schwartz, 1993).
- 103 4. In order to capture regional climatic effects we calculated the mean spring temperature by using the
104 daily mean temperature from March 1 through May 30.
- 105 5. Mean spring temperature was calculated – likely after chilling was accumulated – in an attempt to
106 incorporate the general effects of spring forcing temperatures in our Bayesian hierarchical model and
107 to compare differences in spring across sites (Basler & Körner, 2012; Körner *et al.*, 2016).
- 108 6. We collected NAO-index data from the KNMI Climate Explorer annual NAO time series and selected
109 the NAO indices from November until April to best capture the effects of NAO on budburst for each
110 region (Trouet *et al.*, 2009).
- 111 7. Since the primary aim of the study is to predict false spring incidence in a changing climate, we split the
112 data: before temperature trends increased (1951-1983) and after trends increased (1984-2016, Stocker
113 *et al.*, 2013; Kharouba *et al.*, 2018).

114 **Data Analysis**

- 115 1. A false spring was determined if temperatures fell below -2.2°C at least once between budburst and
116 leafout.
- 117 (a) We scaled all of the predictors and found a z-score following the binary predictor approach in order
118 to best compare the effects of each climate variable to each other (Gelman & Hill, 2006).

- 119 (b) We used a space parameter, rather than a more traditional latitude parameter, to adjust for
 120 spatial autocorrelation issues using a minimization of Moran's I of the residuals (David *et al.*,
 121 2017) (Figure S1).
- 122 (c) We then took the calculated eigenvectors determined from the MIR approach and regressed these
 123 against the number of false springs for each datapoint to establish a spatial parameter (space).
- 124 2. We used a Bayesian hierarchical model approach to analyze our data to best estimate the number of
 125 false springs across-species levels.
- 126 (a) We fit a bernoulli distribution model using mean spring temperature, NAO, elevation, distance
 127 from the coast, space, and climate change as predictors and all two-way interactions (fixed effects)
 128 and species as two-way interactions to simulate modeled groups on the main effects.
- 129 (b) The Bayesian hierarchical model was fit using the brms package (Bürkner, Paul-Christian , 2017),
 130 version 2.3.1, in R (R Development Core Team, 2017), version 3.3.1, and was written as follows:
 131 (subject to change as per above, this is just the ideal model)
- $$y_i \sim N(\alpha(i)) + \beta_{MeanSpringTemp_{(i)}} + \beta_{NAO_{(i)}} + \beta_{Elevation_{(i)}} + \beta_{DistanceCoast_{(i)}} + \beta_{Space_{(i)}} \\ + \beta_{ClimateChange_{(i)}} + \beta_{MeanSpringTemp \times Species_{(i)}} + \beta_{NAO \times Species_{(i)}} + \beta_{Elevation \times Species_{(i)}} \\ + \beta_{DistanceCoast \times Species_{(i)}} + \beta_{Space \times Species_{(i)}} + \beta_{ClimateChange \times Species_{(i)}} \\ + \beta_{MeanSpringTemp \times ClimateChange_{(i)}} + \beta_{NAO \times ClimateChange_{(i)}} + \beta_{Elevation \times ClimateChange_{(i)}} \\ + \beta_{DistanceCoast \times ClimateChange_{(i)}} + \beta_{Space \times ClimateChange_{(i)}} + \sigma_{sp(i)}$$
- 132 (c) We ran two chains, each with 4,000 warm-up iterations and 2,500 sampling iterations for a total
 133 of 10,000 posterior samples for each predictor.
- 134 (d) We evaluated our model performance on \hat{R} values that were close to one and assessed chain
 135 convergence and posterior predictive checks (Gelman & Hill, 2006).

133 Results

134 Species variation in budburst and false spring incidence

- 135 1. There is variation in day of budburst across the six species and across space (Figure 1).
- 136 (a) The top three species (*Betula pendula*, *Alnus glutinosa*, *Aesculus hippocastenum*) generally initiated
 137 budburst earlier than the bottom three species (*Fagus sylvatica*, *Quercus robur*, *Fraxinus excelsior*).

- 138 (b) Across all species, certain regions tended to initiate budburst earlier than others (i.e. United
139 Kingdom was earlier than parts of Austria).
- 140 2. All species initiated budburst earlier after 1983 (Figure 2) and the minimum temperature between
141 budburst and leafout was, on average, higher after 1983.
- 142 3. Species that initiated budburst early were not always at a higher risk of false springs (Figure 2), as is
143 evident by *Alnus glutinosa*.
- 144 (a) As seen in Figure (3), mean spring temperature for most species ranged from -5°C to 12°C, but for
145 *Alnus glutinosa* and *Fraxinus excelsior* the mean spring temperature rarely dropped below 0°C.
- 146 (b) The average minimum temperature between budburst and leafout, however, varied across the
147 six species with *Betula pendula* and *Aesculus hippocastanum* experiencing the lowest minimum
148 temperatures.

149 **The effects of climatic regional variation on false spring incidence**

- 150 1. The effects of the predictors varied in both direction and magnitude (Figure 5).
- 151 (a) Mean spring temperature had the biggest effect on the number of false springs, with warmer spring
152 temperatures resulting in fewer false springs.
- 153 (b) More positive NAO indices, higher elevations, and sites further from the coast all increased the
154 likelihood of false springs.
- 155 (c) Overall, there were fewer false springs after 1983.
- 156 2. Most of the interactions with increasing temperatures (i.e. the Climate Change predictor), demonstrate
157 a decreased risk in false springs, however, the rate of false spring risk increased at sites further from
158 the coast after 1983 (Figure 6).

159 **The effects of changing the temperature threshold**

- 160 1. By changing the temperature threshold for defining a false spring — from -2.2°C to -5°C — many of the
161 predictors changed in both magnitude and direction but overall there were fewer false springs (Figure
162 7).

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266 **Tables and Figures**

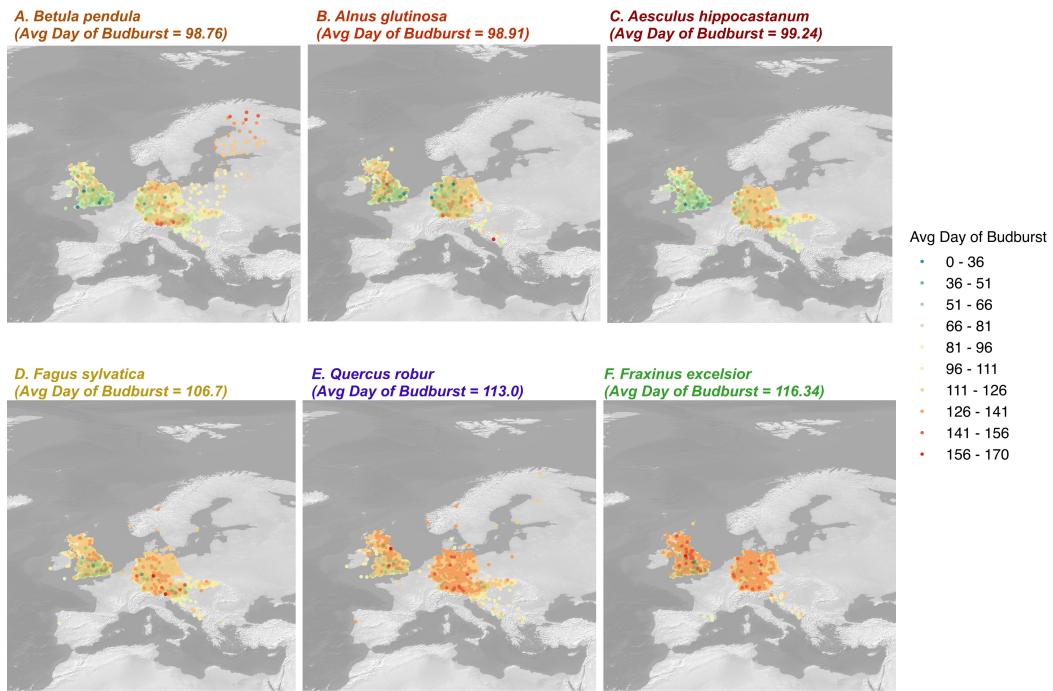


Figure 1: The average day of budburst is mapped by site for each species. Species are ordered by day of budburst starting with *Betula pendula* as the earliest budburst date to *Fraxinus excelsior*. Earlier budburst dates are blue and later budburst dates are in red.

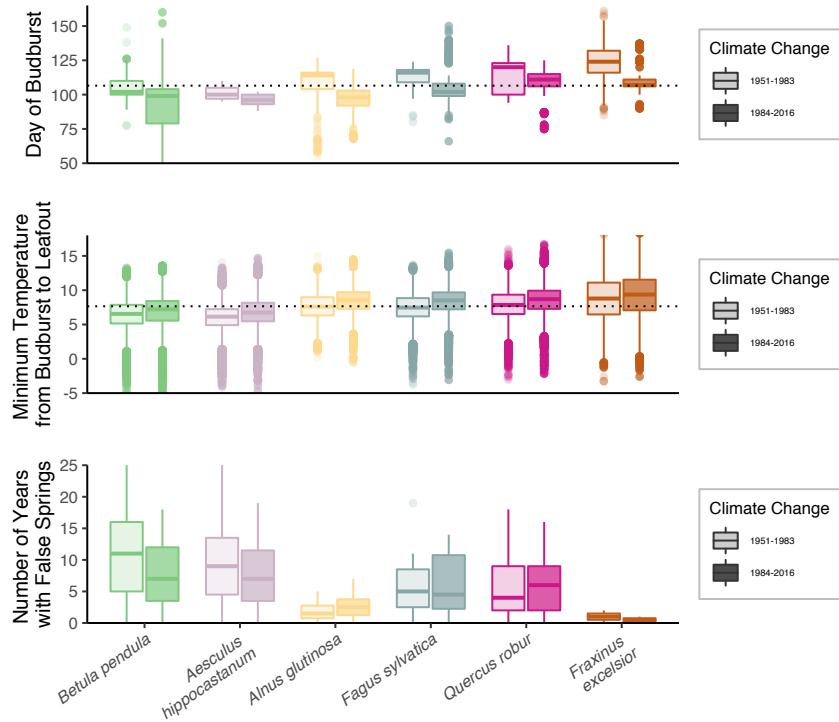


Figure 2

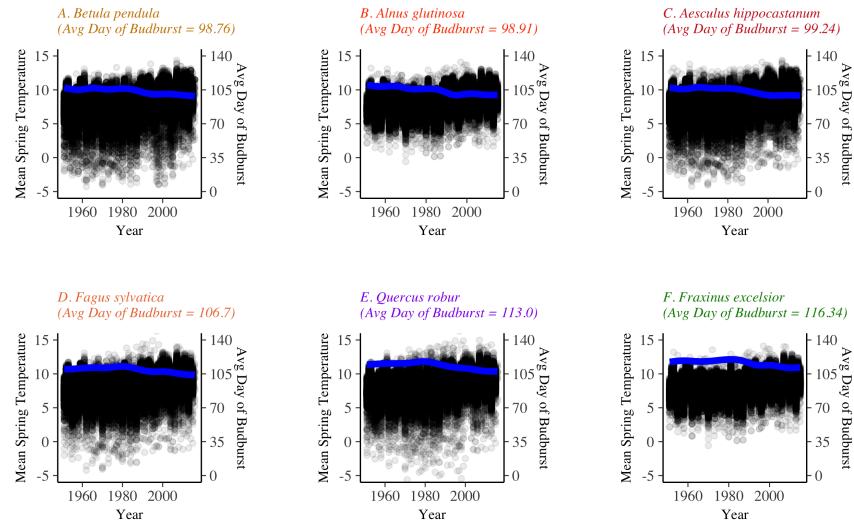


Figure 3

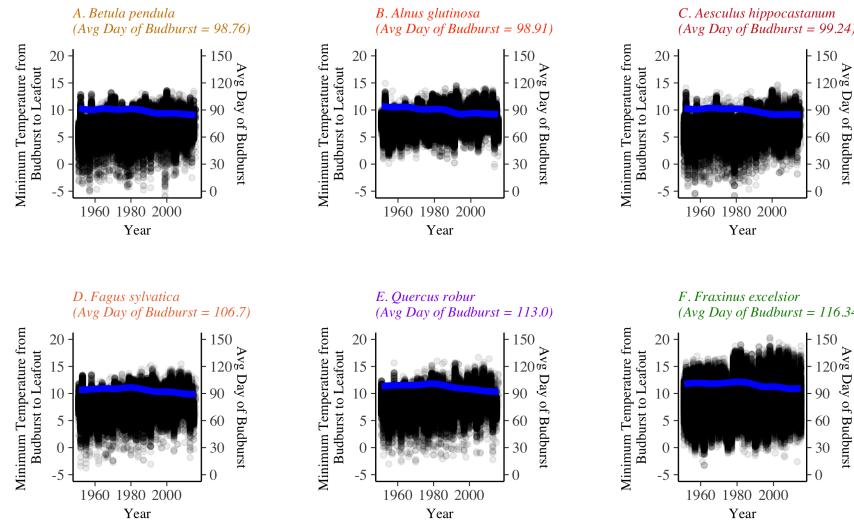


Figure 4

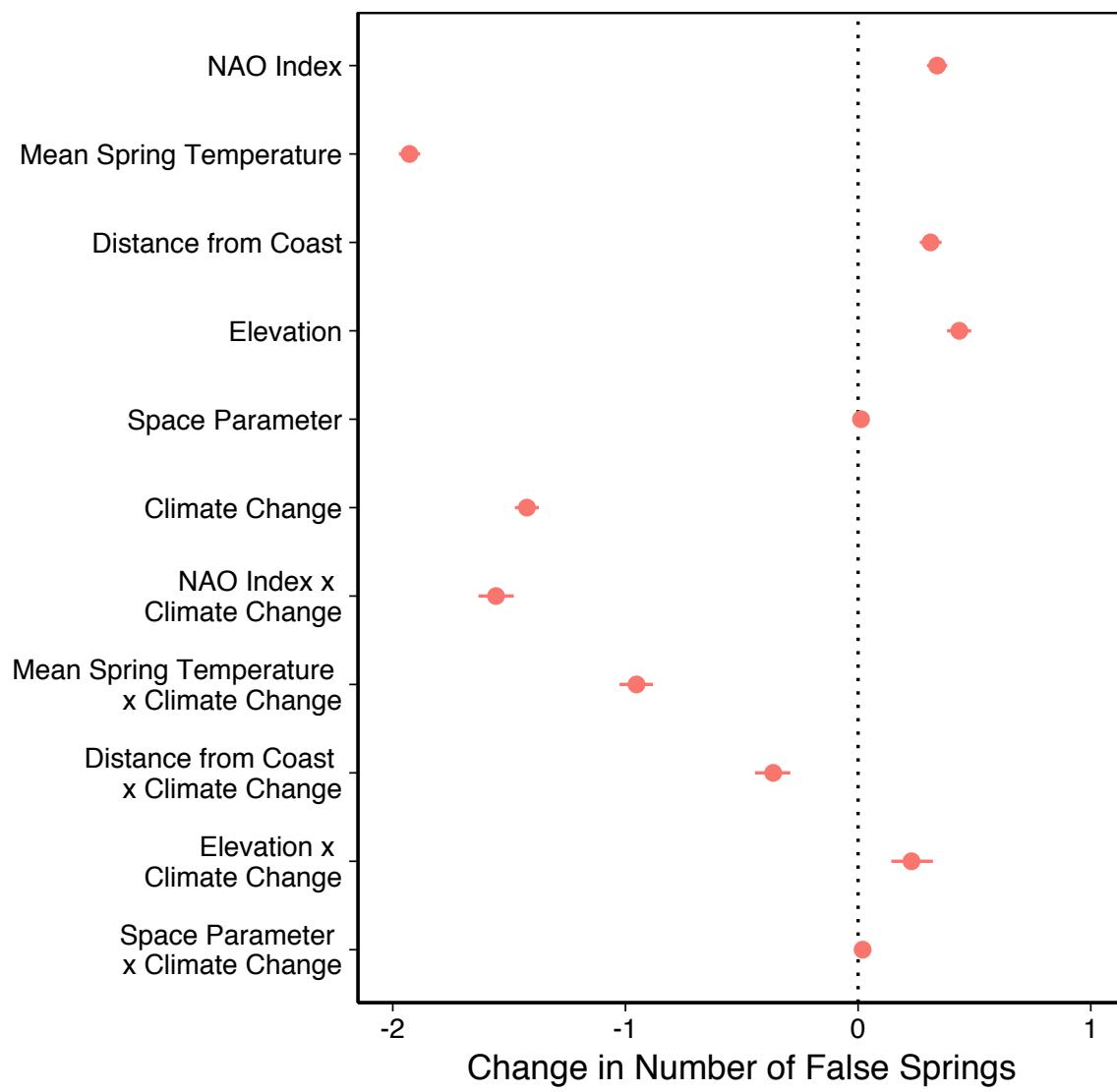


Figure 5

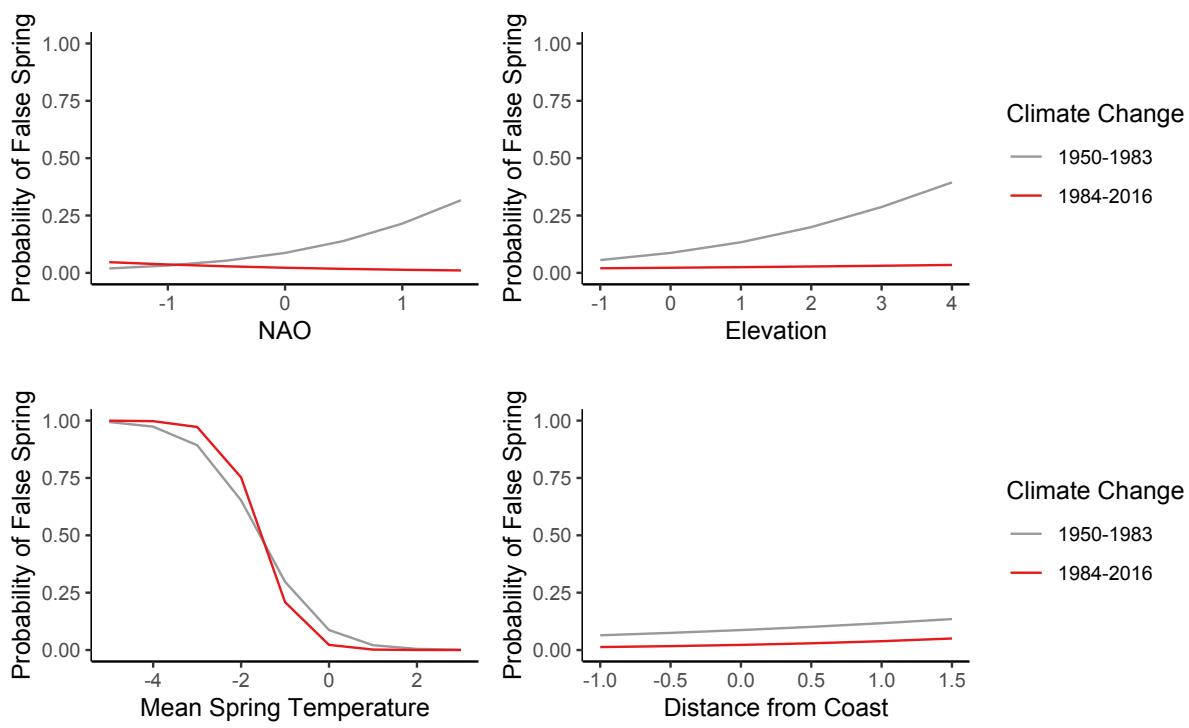


Figure 6

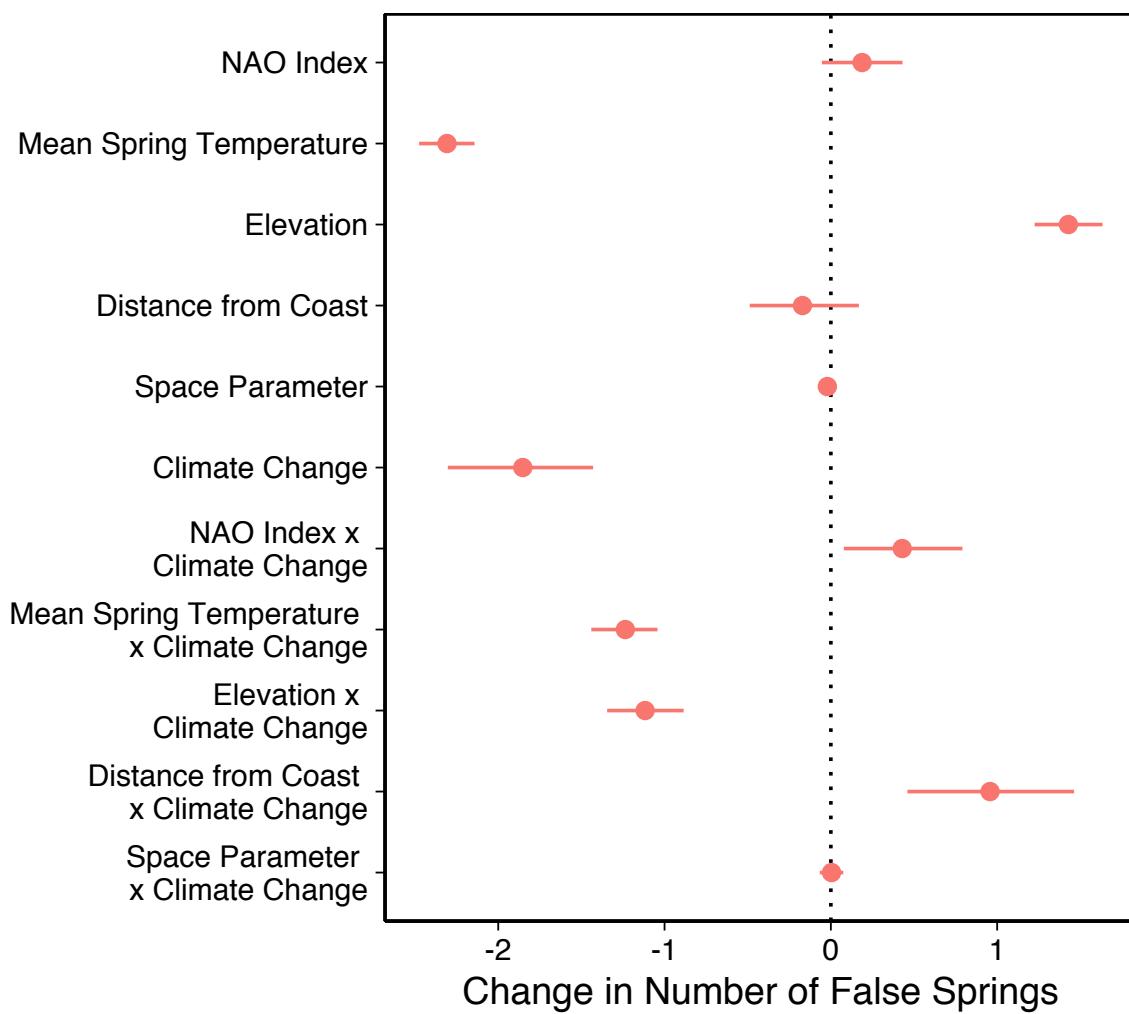


Figure 7

²⁶⁷ **Supplement: Tables and Figures**

²⁶⁸

Table 1: Data points collected for each species

Species	Num. of Observations	Num. of Sites	Num. of Years
<i>Aesculus hippocastanum</i>	156836	10158	66
<i>Alnus glutinosa</i>	91182	6775	66
<i>Betula pendula</i>	155251	10139	66
<i>Fagus sylvatica</i>	129133	9099	66
<i>Fraxinus excelsior</i>	92665	7327	65
<i>Quercus robur</i>	131635	8811	66

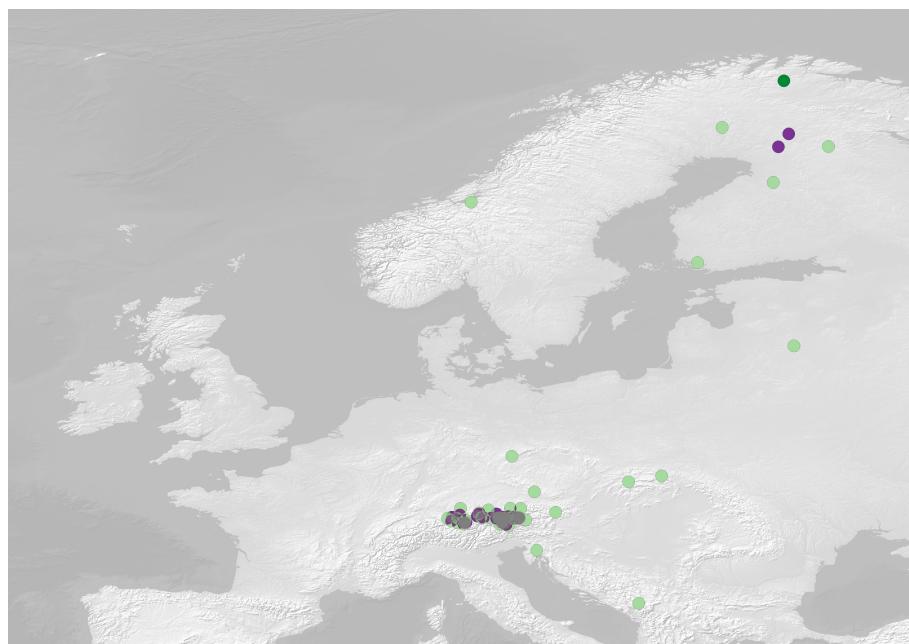


Figure 8: Space parameter values are mapped for each location to elucidate patterns in the parameter values.