

# <sup>1</sup> Regional Risk Outline

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<sup>10</sup>

## <sup>11</sup> Introduction

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

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

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

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

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

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

<sup>28</sup> 2. Episodic frosts are one of the largest limiting factors in species range limits and have shaped plant life  
<sup>29</sup> 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) False spring events can result in photosynthetic tissue loss, which could potentially impact multiple  
39 years of growth and, with the growing season extending, individuals could be exposed to more frosts  
40 in the future (Liu *et al.*, 2018).
- 41 (e) Individuals and species that initiate budburst earlier in the season are more frost resistant (Körner  
42 *et al.*, 2016), however, as climate change advances, less frost resistant individuals may start initiating  
43 budburst before the last freeze date.
- 44 (f) Despite the importance of false spring events, the extent of damage and the frequency and intensity  
45 of false spring events is still largely unknown.
- 46 3. The North Atlantic Oscillation (NAO) index is often used to describe winter and spring circulation  
47 across Europe.
- 48 (a) More positive NAO phases tend to result in higher than average winter and early spring tempera-  
49 tures, and with climate change, higher NAO phases has correlated to even earlier budburst dates  
50 in some regions (Chmielewski & Rötzer, 2001), however it is unclear if more positive NAO phases  
51 also translates to more false springs.
- 52 4. There is large debate over whether or not spring freeze damage will increase (Hänninen, 1991; Augspurger,  
53 2013; Labe *et al.*, 2016), remain the same (Scheifinger *et al.*, 2003) or even decrease (Kramer, 1994;  
54 Vitra *et al.*, 2017) with climate change and there is also great variation within studies.
- 55 (a) Some research suggests false spring incidence has declined in many regions (i.e. across parts of  
56 North America and Asia), however the prevalence of spring frosts has consistently increased across  
57 Europe since 1982 (Liu *et al.*, 2018).
- 58 (b) However, recent studies have demonstrated regional effects may be more closely related to false  
59 spring risk: whether via altitudinal variation (Vitra *et al.*, 2017) or distance from the coast  
60 (Wypych *et al.*, 2016).

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

## 84 Methods

### 85 Phenological Data and Calculating Vegetative Risk

- 86 1. We obtained phenological data from the Pan European Phenology network (PEP725, [www.pep725.edu](http://www.pep725.edu)),  
87 which provides open access phenology records across Europe (Templ *et al.*, 2018).
- 88 2. Since plants are most susceptible to damage from frost between budburst and full leafout, we selected  
89 only leafout data (i.e., in Meier, 2001, BBCH 11, which is defined as the point of leaf unfolding and the

- 90 first visible leaf stalk) from the PEP725 dataset.
- 91 3. We then subtracted 12 days from the leafout date to establish a rough estimate for day of budburst  
92 (Donnelly *et al.*, 2017).
- 93 4. The species used in the study were *Aesculus hippocastanum* L., *Alnus glutinosa* (L.) Gaertn., *Betula*  
94 *pendula* Roth., *Fagus sylvatica* Ehrh., *Fraxinus excelsior* L., *Quercus robur* L.
- 95 5. Selection criteria for the species were as follows: (1) to be temperate, deciduous species that were not  
96 cultivars or used for crops, (2) there were at least 140,000 observations of BBCH 11, (3) to represent  
97 over half of the total number of sites available (11,684), (4) there were observations for at least 65 out  
98 of the 66 years of the study (1951-2016).

## 99 Climate Data

- 100 1. We collected daily gridded climate data from the European Climate Assessment & Dataset (ECA&D)  
101 and used the E-OBS 0.25 degree regular latitude-longitude grid from version 16.
- 102 2. We used the daily minimum temperature dataset to determine if a false spring occurred.
- 103 3. False springs in this study were defined as temperatures at or below -2.2°C (Schwartz, 1993).
- 104 4. In order to capture regional climatic effects we calculated the mean spring temperature by using the  
105 daily mean temperature from February 1 through April 30.
- 106 5. Mean spring temperature was calculated – likely after chilling was accumulated – in an attempt to  
107 incorporate the general effects of spring forcing temperatures in our Bayesian hierarchical model and  
108 to compare differences in spring across sites (Basler & Körner, 2012; Körner *et al.*, 2016).
- 109 6. We collected NAO-index data from the KNMI Climate Explorer annual NAO time series (Trouet *et al.*,  
110 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 the elevation predictor by dividing it by 100 to be consistent with the other predictors  
 118 in the model.
- 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 mixed-effects model using mean spring temperature, NAO, elevation, space, and climate  
 127 change as predictors and all two-way interactions (fixed effects) and species as modeled groups on  
 128 the main effects (random effects).
- 129 (b) (Although, we may find that space and elevation are collinear and may decide to take space out  
 130 and use elevation instead - unless space is picking up more of the distance from the coast then  
 131 we'll use the space parameter instead of elevation)
- 132 (c) The Bayesian hierarchical model was fit using the brms package (Bürkner, Paul-Christian , 2017),  
 version 2.3.1, in R (R Development Core Team, 2017), version 3.3.1, and was written as follows:  
 (subject to change as per above, this is just the ideal model)

$$y_i \sim N(\alpha(i) + \beta_{MeanSpringTemp_{sp(i)}} + \beta_{NAO_{sp(i)}} + \beta_{elevation_{sp(i)}} + \beta_{space_{sp(i)}} + \beta_{ClimateChange_{sp(i)}} \\ + \beta_{MeanSpringTemp \times ClimateChange_{(i)}} + \beta_{NAO \times ClimateChange_{(i)}} \\ + \beta_{elevation \times ClimateChange_{(i)}} + \beta_{space \times ClimateChange_{(i)}} + \sigma_{sp(i)})$$

- (d) The  $\beta$  coefficients and  $\alpha$  were modeled at the species level:

$$1. \beta_{MeanSpringTemp_{sp}} \sim N(\mu_{MeanSpringTemp}, \sigma^2_{MeanSpringTemp}) \\ \dots \\ 5. \beta_{ClimateChange_{sp}} \sim N(\mu_{ClimateChange}, \sigma^2_{ClimateChange})$$

- 132 (e) 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 (f) We evaluated our model performance on  $\hat{R}$  values that were close to 1 and assessed chain conver-  
 135 gence and posterior predictive checks (Gelman & Hill, 2006).

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234 **Tables and Figures**

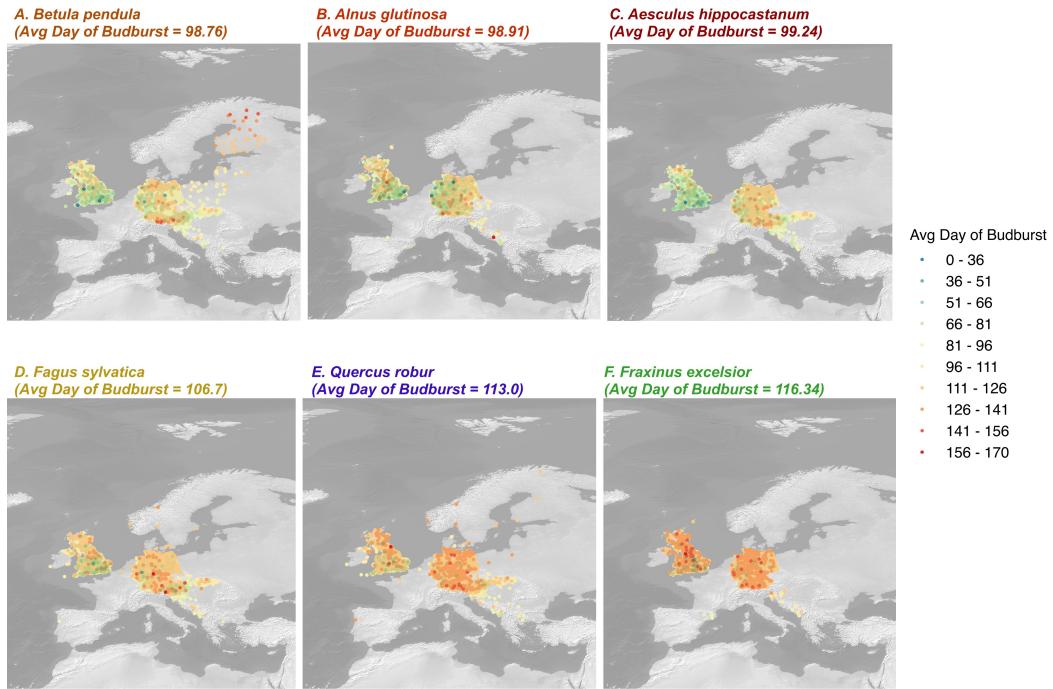


Figure 1: The average day of budburst is mapped by site for each species. Species are ordered by day of budburst starting with *Betula pendula* as the earliest budburst date to *Fraxinus excelsior*. Earlier budburst dates are blue and later budburst dates are in red.

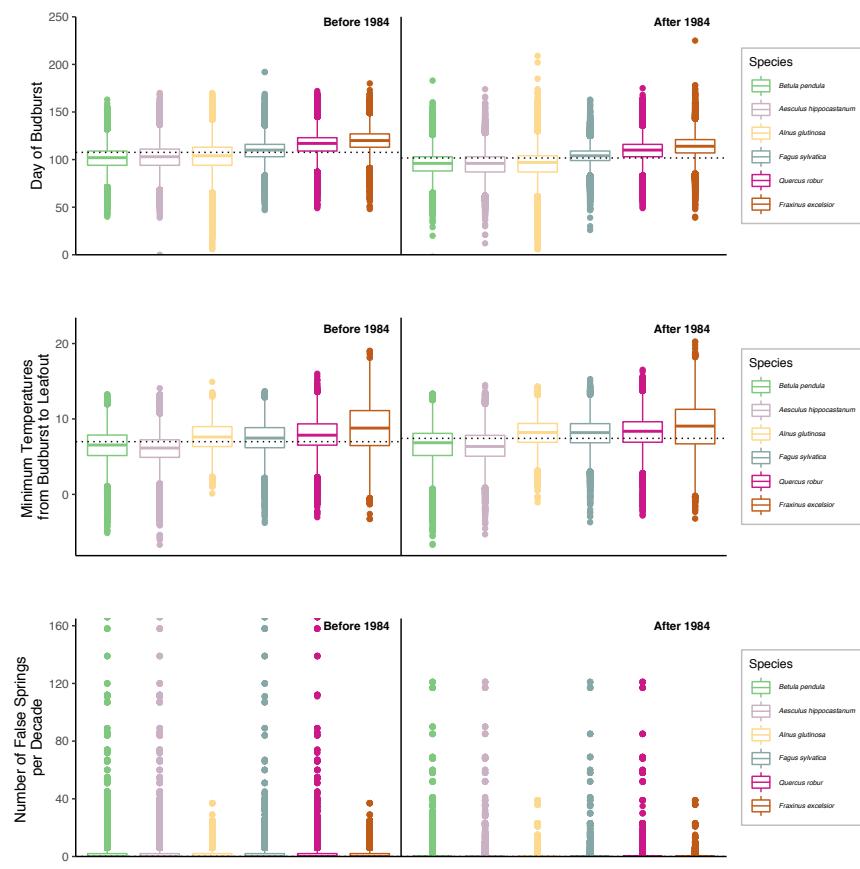


Figure 2

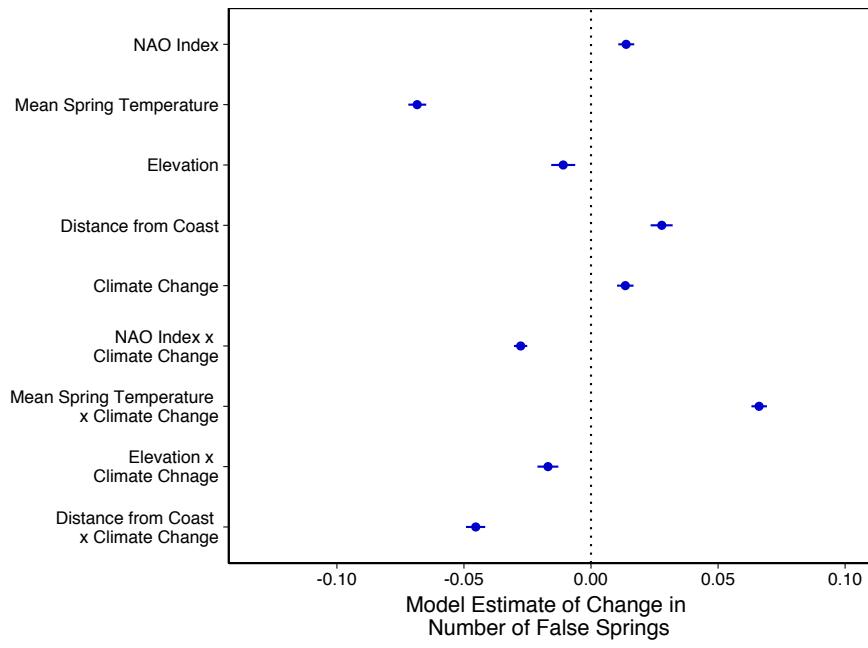


Figure 3

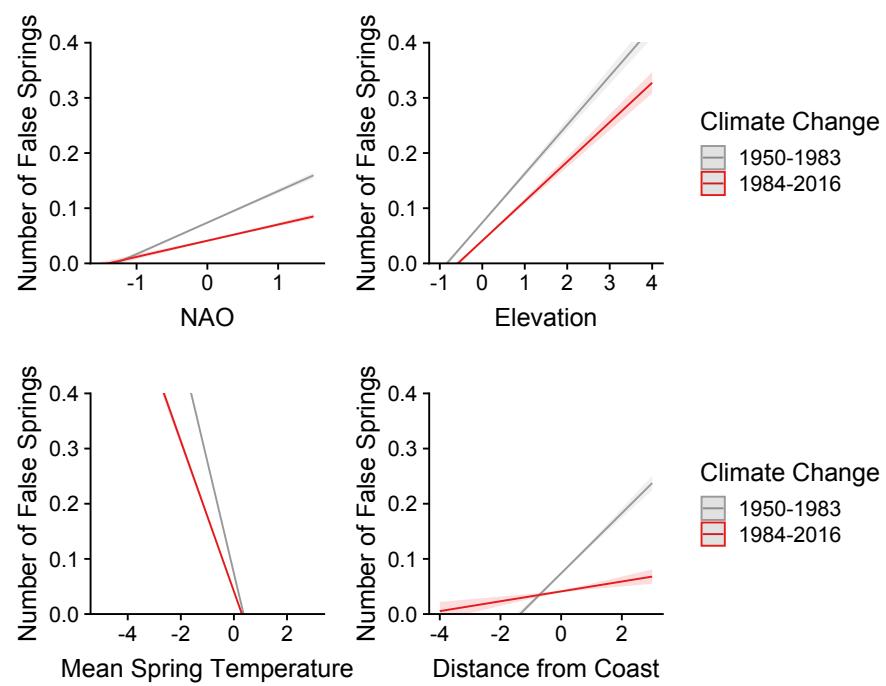


Figure 4

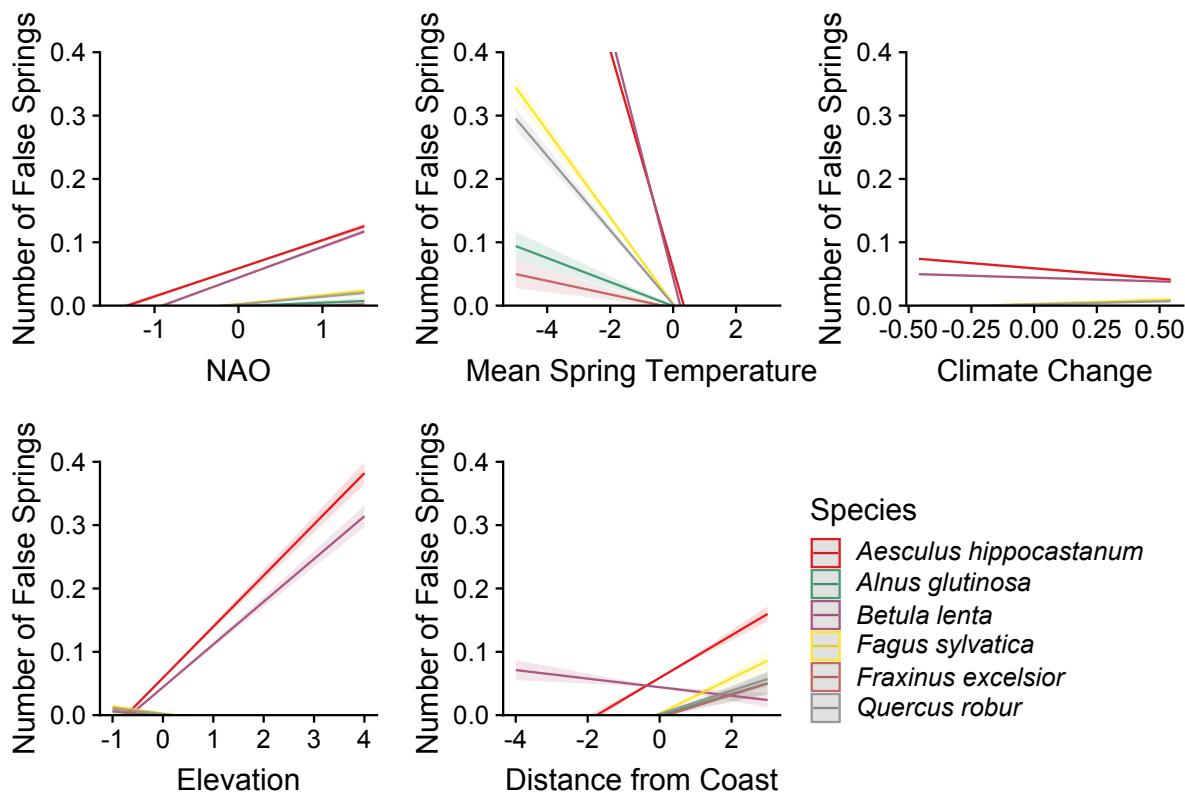


Figure 5

## 235 Supplement: Tables and Figures

236

Table 1: Data points collected for each species

237

Species	Num. of Observations	Num. of Sites	Num. of Years
<i>Aesculus hippocastanum</i>	216396	10158	66
<i>Alnus glutinosa</i>	136991	6775	66
<i>Betula pendula</i>	215729	10139	66
<i>Fagus sylvatica</i>	185949	9099	66
<i>Fraxinus excelsior</i>	143269	7327	65
<i>Quercus robur</i>	184406	8811	66

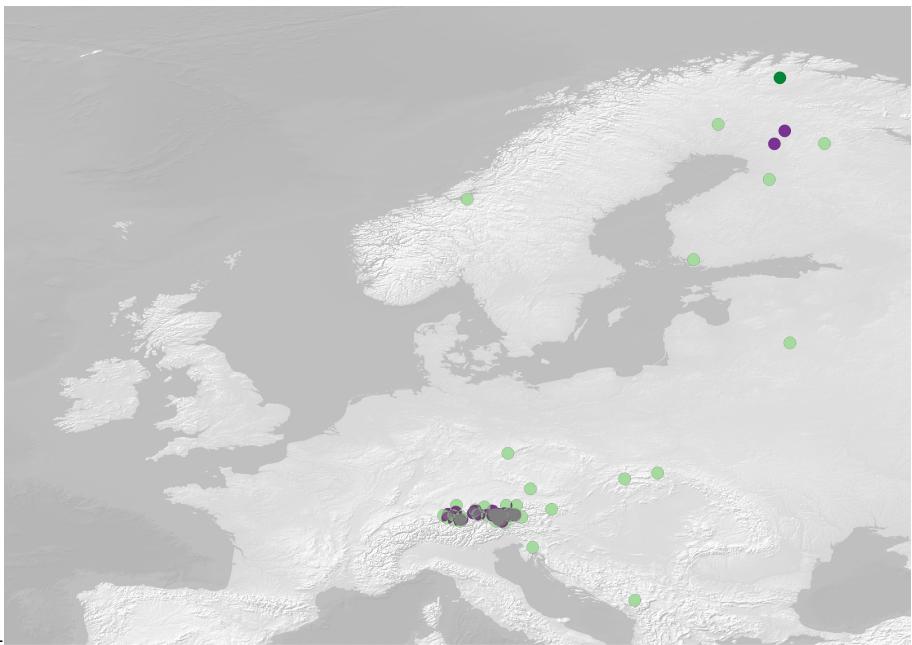


Figure 6: Space parameter values are mapped for each location to elucidate patterns in the parameter values.