

Climate change reshapes the drivers of false spring risk across European trees

Authors:

C. J. Chamberlain^{1,2}, B. I. Cook³, I. Morales-Castilla^{4,5} & E. M. Wolkovich^{1,2,6}

Author affiliations:

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

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

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

⁴GloCEE - Global Change Ecology and Evolution Group, Department of Life Sciences, Universidad de Alcalá, Alcalá de Henares, 28805, Spain

⁵Department of Environmental Science and Policy, George Mason University, Fairfax, VA 22030;

⁶Forest & Conservation Sciences, Faculty of Forestry, University of British Columbia, 2424 Main Mall, Vancouver, BC V6T 1Z4

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

Methods: Spatial predictor

Spatial autocorrelation (SA) is a common issue in spatial ecology given that nearby spatial units tend to be more similar than units far apart, and thus, cannot be considered as independent units, which is a frequent assumption in statistical tests (Diniz-Filho *et al.*, 2003). If model residuals are spatially autocorrelated, and thus, non-independent then model coefficients and errors may be biased in a hard-to-predict way (Mauricio Bini *et al.*, 2009). On the contrary, if model residuals are not autocorrelated, then SA should not be of concern (Hawkins, 2012).

To control for spatial autocorrelation and to account for spatially structured processes independent from our environmental predictors of false springs, we generated an additional spatial predictor for the model. To avoid collinearity, we computed our spatial predictor from the residuals of a linear model of false springs as a function (Equation S1) of all other factors that are also spatially structured (e.g. spring temperature, altitude, distance to the coast), following the logic of spatial filter modelling (Diniz-Filho & Bini, 2005). The calculation of the spatial predictor followed the next steps: (a) we fit a linear model of false spring versus environmental factors, (b) we extracted the residuals of the regression Equation S1, which represent

the portion of the variation in the number of false springs that is independent from the predictors in the model and (c) we utilized the residuals as our y_i values in a selection of spatial eigenvectors to retain only the minimal subset of spatial eigenvectors that are able to remove SA from model residuals. Specifically, we selected eigenvectors following the the minimization of Moran's I of the residuals (MIR) approach (Griffith & Peres-Neto, 2006; Diniz-Filho *et al.*, 2012; Bauman *et al.*, 2017). (d) Next, we fit a linear model between the residuals of Equation S1 and the subset of selected eigenvectors. And, finally, (e) we took the fitted values from this regression as our spatial predictor in our final model (see equation from main text, Equation 1), which can be interpreted as a latent variable summarizing the spatial structure in false springs that is unaccounted for by the rest of the environmental factors in our model (Morales-Castilla *et al.*, 2012). A spatial predictor generated in this way has three major advantages. First, it ensures that no SA is left in model residuals. Second, it avoids introducing collinearity issues with other predictors in the model. And third, it can be interpreted as a latent variable summarizing spatial processes (e.g. local adaptation, plasticity, etc.) occurring at multiple scales.

$$\begin{aligned}
y_i = & \alpha_{[i]} + \beta_{NAO_{[i]}} + \beta_{MST_{[i]}} + \beta_{Elevation_{[i]}} + \beta_{DistanceCoast_{[i]}} \\
& + \beta_{ClimateChange_{[i]}} + \beta_{NAO \times Species_{[i]}} + \beta_{MST \times Species_{[i]}} + \beta_{Elevation \times Species_{[i]}} \\
& + \beta_{DistanceCoast \times Species_{[i]}} + \beta_{ClimateChange \times Species_{[i]}} \\
& + \beta_{NAO \times ClimateChange_{[i]}} + \beta_{MST \times ClimateChange_{[i]}} + \beta_{Elevation \times ClimateChange_{[i]}} \\
& + \beta_{DistanceCoast \times ClimateChange_{[i]}} + \sigma_{[i]}
\end{aligned} \tag{S1}$$

Species rate of budburst calculations

Due to the paucity of data for BBCH 7 in the PEP725 dataset, we were unable to use observations for both budburst and leafout to determine the durations of vegetative risk. Instead, we used data from a growth chamber experiment (Flynn & Wolkovich, 2018) to determine the average number of days between budburst and leafout for our study species. We took the mean number of days between budburst and leafout for the entire experiment, which was 12 days. We compared this number to a field observation study (Donnelly *et al.*, 2017) that looked at the time between budburst and leafout across 10 species over 5 years. Finally, we assessed data that were provided by the USA National Phenology Network and the many participants who contribute to its Nature's Notebook program (USA-NPN, 2019; www.usanpn.org/data/observational) for *Aesculus flava* (Sol.), *Aesculus glabra* (Willd.), *Alnus incana* (Moench.), *Betula nigra* (L.), *Betula papyrifera* (Marshall), *Fagus grandifolia* (Ehrh.), *Fraxinus americana* (L.), *Fraxinus nigra* (Marshall) and *Quercus velutina* (Lam.) and took the mean number of days between budburst and leafout. Across all three approaches, the average

number of days between budburst and leafout was approximately 12 days.

Again, due to a lack of BBCH 7 data, we were unable to determine species-specific averages of number of days between budburst and leafout. We used a similar approach as above by using data from the growth chamber experiment (Flynn & Wolkovich, 2018) but instead of finding whole experiment means we determined species-specific averages. We used the rate of budburst of *Acer saccharum* (Marshall) for *Aesculus hippocastanum* (Buerki *et al.*, 2010), *Alnus incana* for *Alnus glutinosa*, *Betula papyrifera* for *Betula pendula* (Wang *et al.*, 2016), *Fagus grandifolia* for *Fagus sylvatica*, *Fraxinus nigra* for *Fraxinus excelsior* and *Quercus alba* (L.) for *Quercus robur* (Hipp *et al.*, 2017).

Results: The effects of climatic and spatial variation on false spring incidence

Most species had mean spring temperatures that ranged from -5°C to 12°C, but for *Alnus glutinosa* and *Fraxinus excelsior* temperatures rarely dropped below 0°C, whereas *Quercus robur* experienced some of the lowest spring temperatures (see Figure S1).

The overall model output estimates are for *Aesculus hippocastanum* as species were used as two-way interactions to simulate modeled groups on the main effects. For the model estimate of change plots (Figure 3, Figure S3 and Figure S4), we used average predictive comparisons to calculate the mean estimates and the posteriors for each species and then took the average to find the overall effect for each predictor. The model estimates on the logit scale were converted to probability percentages for easier interpretation by using the ‘divide by 4’ rule (Gelman & Hill, 2006) and then back converted to the original scale.

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Supplement: Tables and Figures

Table S1: Data collected from PEP725 for each species and the calculated number of false spring years

Species	Num. of Observations	Num. of False Springs	Num. of Sites	Num. of Years
<i>Aesculus hippocastanum</i>	156468	44746	10157	66
<i>Alnus glutinosa</i>	91094	27296	6775	65
<i>Betula pendula</i>	154897	46685	10139	66
<i>Fagus sylvatica</i>	129133	29237	9099	66
<i>Fraxinus excelsior</i>	92665	8256	7327	65
<i>Quercus robur</i>	131635	16657	8811	66

Table S2: Mean day of budburst and standard deviation for each species for before (1951-1983) and after climate change (1984-2016).

	1951-1983		1984-2016	
	mean	sd	mean	sd
<i>Aesculus hippocastanum</i>	102.2	12.44	95.35	12.09
<i>Alnus glutinosa</i>	102.8	14.81	94.90	14.71
<i>Betula pendula</i>	101.3	11.76	95.44	11.25
<i>Fagus sylvatica</i>	109.1	9.978	103.7	9.623
<i>Fraxinus excelsior</i>	119.4	11.79	113.5	11.53
<i>Quercus robur</i>	115.9	11.31	109.6	10.95

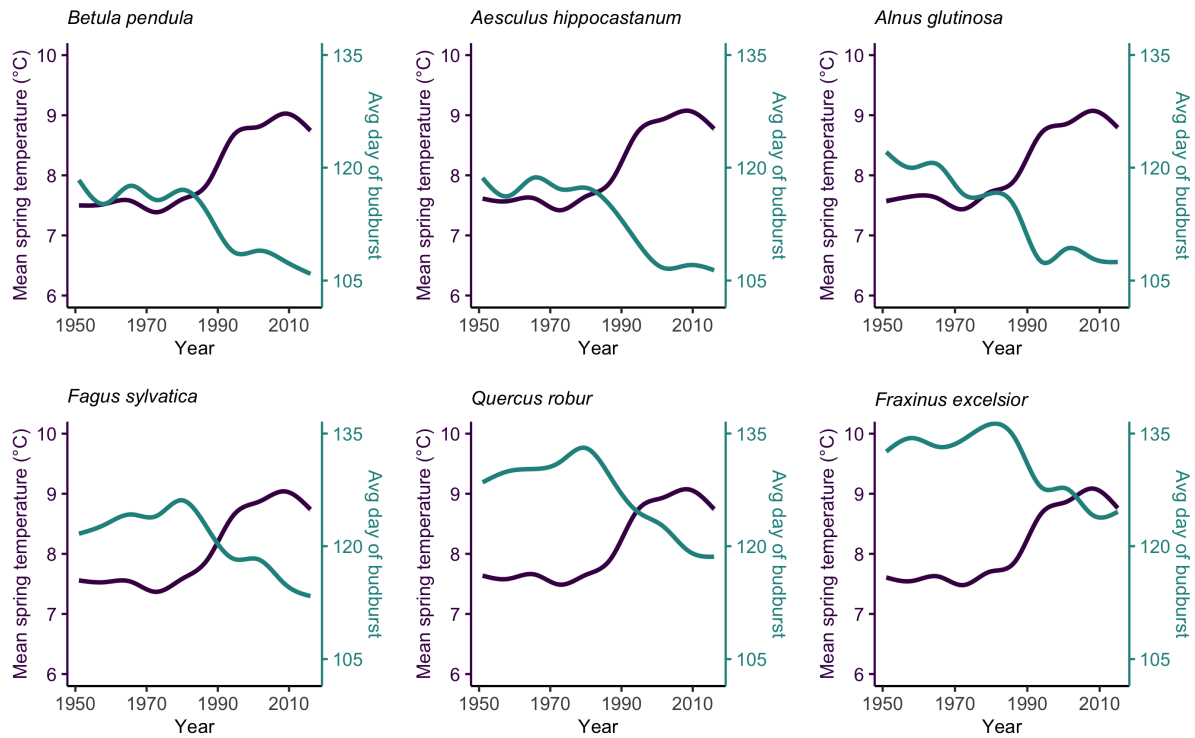


Figure S1: Mean spring temperatures are plotted for each site and year (from 1951-2016) for each species. The purple line shows the trend in mean spring temperatures from March 1 to May 31 and the green line represents the trend of average day of budburst for each year for each species. Both lines are cyclic penalized cubic regression spline smooths with basis dimensions equal to the number of years in the study (i.e., 66). Species are ordered by average day of budburst, with the earliest being *Betula pendula* and the latest being *Fraxinus excelsior*.

```
## Error in readChar(con, 5L, useBytes = TRUE): cannot open the connection
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## Error in tidy(bb.mod, prob = 0.9): object 'bb.mod' not found

## Error in gsub("b_", "", bb$term): object 'bb' not found

## Error in gsub("Intercept", "Intercept", bb$term): object 'bb' not found

## Error in gsub("cc", "Climate Change", bb$term): object 'bb' not found

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## Error in gsub("speciesBETPEN", "Betula pendula", bb$term): object 'bb' not found

## Error in gsub("speciesFAGSYL", "Fagus sylvatica", bb$term): object 'bb' not found

## Error in gsub("speciesFRAEXC", "Fraxinus excelsior", bb$term): object 'bb' not found

## Error in gsub("speciesQUEROB", "Quercus robur", bb$term): object 'bb' not found

## Error in gsub(":", " by ", bb$term): object 'bb' not found

## Error in colnames(bb) <- c("", "Estimate", "std.err", "10\\%", "90\\%"): object 'bb' not
found

## Error in xtable(bb[1:12, c(1:2, 4:5)], caption = "Summary of simple linear regression model
of day of budburst before and after climate change across species.", : object 'bb' not found
```


Table S3: Summary of simple linear regression model of average minimum temperature between budburst and leafout before and after climate change across species.

	Estimate	10%	90%
Intercept	5.99	5.97	6.00
Climate Change	0.83	0.81	0.85
<i>Alnus glutinosa</i>	1.65	1.63	1.67
<i>Betula pendula</i>	0.50	0.48	0.52
<i>Fagus sylvatica</i>	1.50	1.48	1.51
<i>Fraxinus excelsior</i>	2.76	2.74	2.78
<i>Quercus robur</i>	1.88	1.87	1.90
Climate Change by <i>Alnus glutinosa</i>	-0.04	-0.07	-0.01
Climate Change by <i>Betula pendula</i>	-0.31	-0.34	-0.29
Climate Change by <i>Fagus sylvatica</i>	0.08	0.05	0.10
Climate Change by <i>Fraxinus excelsior</i>	-0.26	-0.29	-0.23
Climate Change by <i>Quercus robur</i>	-0.10	-0.13	-0.07

Table S4: Summary of simple linear regression model of probability of false spring before and after climate change across species.

	Estimate	10%	90%
Intercept	0.27	0.27	0.27
Climate Change	0.04	0.03	0.04
<i>Alnus glutinosa</i>	0.00	-0.00	0.00
<i>Betula pendula</i>	0.01	0.01	0.02
<i>Fagus sylvatica</i>	-0.03	-0.04	-0.03
<i>Fraxinus excelsior</i>	-0.17	-0.17	-0.17
<i>Quercus robur</i>	-0.14	-0.14	-0.13
Climate Change by <i>Alnus glutinosa</i>	0.02	0.01	0.03
Climate Change by <i>Betula pendula</i>	0.00	-0.00	0.01
Climate Change by <i>Fagus sylvatica</i>	-0.06	-0.06	-0.05
Climate Change by <i>Fraxinus excelsior</i>	-0.06	-0.06	-0.05
Climate Change by <i>Quercus robur</i>	-0.05	-0.06	-0.05

Table S5: Summary of Bernoulli model of false spring risk with varying rates of budburst for each species without the species interactions (estimates presented on logit scale for *Aesculus hippocastanum* as the intercept).

	Estimate	10%	25%	75%	90%
Intercept	-0.88	-0.89	-0.89	-0.88	-0.88
NAO Index	0.14	0.12	0.13	0.15	0.16
Mean Spring Temperature	-0.48	-0.50	-0.49	-0.47	-0.45
Distance from Coast	0.40	0.38	0.39	0.41	0.43
Elevation	0.19	0.16	0.18	0.20	0.22
Space Parameter	-0.06	-0.08	-0.07	-0.06	-0.05
Climate Change	0.35	0.33	0.34	0.36	0.37
<i>Alnus glutinosa</i>	0.05	0.04	0.05	0.06	0.07
<i>Betula pendula</i>	0.06	0.04	0.05	0.06	0.07
<i>Fagus sylvatica</i>	-0.35	-0.37	-0.36	-0.35	-0.34
<i>Fraxinus excelsior</i>	-1.43	-1.45	-1.44	-1.42	-1.41
<i>Quercus robur</i>	-1.03	-1.04	-1.03	-1.02	-1.01
NAO Index by <i>Alnus glutinosa</i>	-0.07	-0.11	-0.09	-0.06	-0.04
NAO Index by <i>Betula pendula</i>	-0.08	-0.11	-0.10	-0.07	-0.05
NAO Index by <i>Fagus sylvatica</i>	0.12	0.09	0.11	0.13	0.15
NAO Index by <i>Fraxinus excelsior</i>	0.03	-0.02	0.01	0.05	0.08
NAO Index by <i>Quercus robur</i>	0.08	0.04	0.06	0.09	0.12
Mean Spring Temperature by <i>Alnus glutinosa</i>	0.14	0.10	0.13	0.16	0.18
Mean Spring Temperature by <i>Betula pendula</i>	-0.02	-0.05	-0.03	-0.01	0.01
Mean Spring Temperature by <i>Fagus sylvatica</i>	-0.06	-0.10	-0.08	-0.05	-0.02
Mean Spring Temperature by <i>Fraxinus excelsior</i>	0.32	0.26	0.30	0.34	0.38
Mean Spring Temperature by <i>Quercus robur</i>	0.10	0.06	0.09	0.12	0.15
Distance from Coast by <i>Alnus glutinosa</i>	-0.07	-0.11	-0.08	-0.05	-0.03
Distance from Coast by <i>Betula pendula</i>	0.01	-0.02	-0.00	0.03	0.05
Distance from Coast by <i>Fagus sylvatica</i>	-0.00	-0.04	-0.02	0.01	0.04
Distance from Coast by <i>Fraxinus excelsior</i>	0.26	0.20	0.23	0.28	0.31
Distance from Coast by <i>Quercus robur</i>	-0.10	-0.14	-0.12	-0.08	-0.05
Elevation by <i>Alnus glutinosa</i>	0.02	-0.02	0.00	0.04	0.07
Elevation by <i>Betula pendula</i>	0.03	-0.01	0.02	0.05	0.07
Elevation by <i>Fagus sylvatica</i>	0.00	-0.04	-0.02	0.02	0.04

Table S5: Summary of Bernoulli model of false spring risk with varying rates of budburst for each species without the species interactions (estimates presented on logit scale for *Aesculus hippocastanum* as the intercept).

	Estimate	10%	25%	75%	90%
Elevation by Fraxinus excelsior	-0.37	-0.43	-0.39	-0.34	-0.31
Elevation by Quercus robur	-0.08	-0.13	-0.10	-0.06	-0.03
Space Parameter by Alnus glutinosa	0.01	-0.02	0.00	0.03	0.05
Space Parameter by Betula pendula	-0.01	-0.03	-0.02	0.00	0.02
Space Parameter by Fagus sylvatica	-0.04	-0.07	-0.05	-0.03	-0.01
Space Parameter by Fraxinus excelsior	0.05	0.01	0.03	0.07	0.09
Space Parameter by Quercus robur	0.02	-0.02	0.00	0.03	0.05
Climate Change by Alnus glutinosa	0.06	0.03	0.05	0.07	0.09
Climate Change by Betula pendula	0.06	0.03	0.04	0.07	0.09
Climate Change by Fagus sylvatica	-0.32	-0.35	-0.34	-0.31	-0.29
Climate Change by Fraxinus excelsior	-0.52	-0.57	-0.54	-0.50	-0.47
Climate Change by Quercus robur	-0.41	-0.44	-0.42	-0.39	-0.37
NAO Index by Climate Change	-0.83	-0.85	-0.84	-0.82	-0.81
Mean Spring Temperature by Climate Change	0.42	0.40	0.41	0.43	0.44
Distance from Coast by Climate Change	-0.12	-0.15	-0.13	-0.11	-0.10
Elevation by Climate Change	-0.00	-0.03	-0.01	0.01	0.03
Space Parameter by Climate Change	-0.05	-0.07	-0.05	-0.04	-0.03

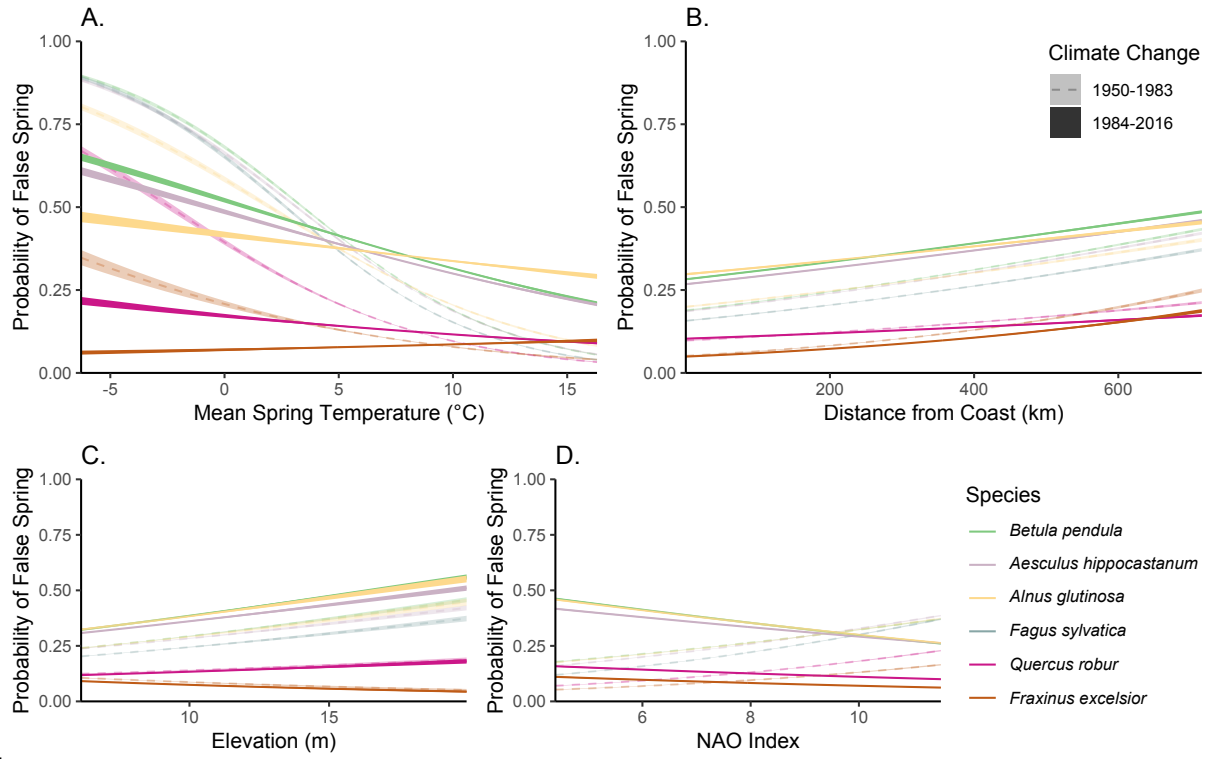


Figure S2: Average predictive comparisons for all climate change interactions with each of the main effects (i.e., mean spring temperature, distance from the coast, elevation, and NAO index) for all species.

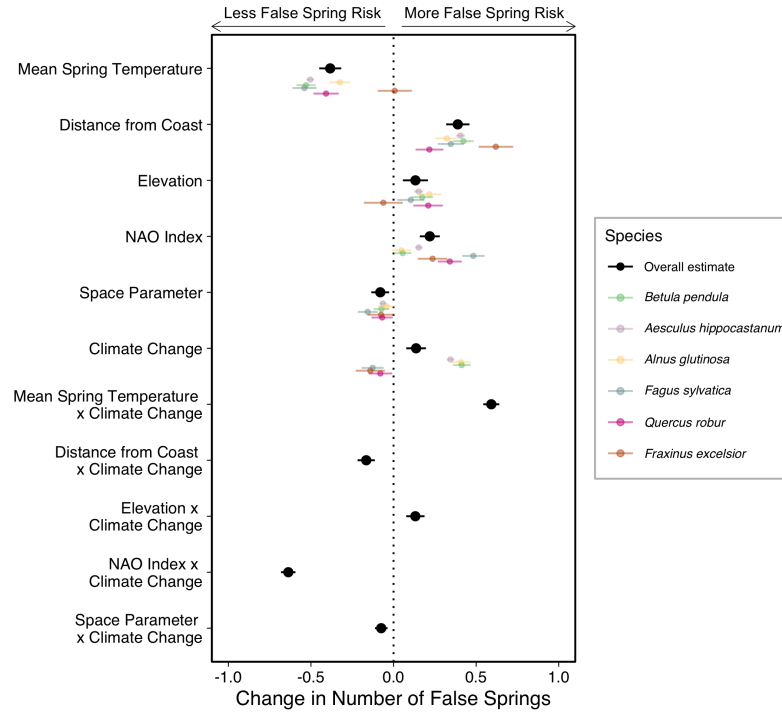


Figure S3: Effects of species, climatic and geographical predictors on false spring risk with different rates of leafout for each species. More positive values indicate an increased probability of a false spring whereas more negative values suggest a lower probability of a false spring. Dots and lines show means and 90% uncertainty intervals. Values closer to zero have less of an effect on false springs. There were 622,565 zeros and 132,463 ones for false springs in the data.

Table S6: Summary of Bernoulli model of false spring risk with varying rates of leafout for each species without the species interactions (estimates presented on logit scale for *Aesculus hippocastanum*).

	Estimate	10%	25%	75%	90%
Intercept	-1.04	-1.05	-1.04	-1.03	-1.03
NAO Index	0.15	0.13	0.14	0.16	0.17
Mean Spring Temperature	-0.50	-0.53	-0.51	-0.49	-0.48
Distance from Coast	0.40	0.38	0.39	0.42	0.43

Table S6: Summary of Bernoulli model of false spring risk with varying rates of leafout for each species without the species interactions (estimates presented on logit scale for *Aesculus hippocastanum*).

	Estimate	10%	25%	75%	90%
Elevation	0.15	0.12	0.14	0.16	0.18
Space Parameter	-0.06	-0.08	-0.07	-0.06	-0.04
Climate Change	0.34	0.32	0.34	0.35	0.37
<i>Alnus glutinosa</i>	0.20	0.18	0.19	0.21	0.22
<i>Betula pendula</i>	0.04	0.03	0.04	0.05	0.06
<i>Fagus sylvatica</i>	-1.38	-1.40	-1.39	-1.37	-1.36
<i>Fraxinus excelsior</i>	-2.11	-2.14	-2.12	-2.10	-2.08
<i>Quercus robur</i>	-1.75	-1.78	-1.76	-1.74	-1.73
NAO Index by <i>Alnus glutinosa</i>	-0.10	-0.14	-0.12	-0.09	-0.07
NAO Index by <i>Betula pendula</i>	-0.10	-0.13	-0.11	-0.08	-0.07
NAO Index by <i>Fagus sylvatica</i>	0.33	0.28	0.31	0.35	0.38
NAO Index by <i>Fraxinus excelsior</i>	0.08	0.02	0.06	0.11	0.15
NAO Index by <i>Quercus robur</i>	0.19	0.14	0.17	0.21	0.24
Mean Spring Temperature by <i>Alnus glutinosa</i>	0.18	0.14	0.16	0.19	0.22
Mean Spring Temperature by <i>Betula pendula</i>	-0.03	-0.06	-0.04	-0.01	0.01
Mean Spring Temperature by <i>Fagus sylvatica</i>	-0.04	-0.08	-0.06	-0.02	0.01
Mean Spring Temperature by <i>Fraxinus excelsior</i>	0.51	0.43	0.48	0.54	0.59
Mean Spring Temperature by <i>Quercus robur</i>	0.10	0.04	0.07	0.12	0.15
Distance from Coast by <i>Alnus glutinosa</i>	-0.08	-0.12	-0.10	-0.07	-0.04
Distance from Coast by <i>Betula pendula</i>	0.02	-0.02	0.00	0.03	0.05
Distance from Coast by <i>Fagus sylvatica</i>	-0.06	-0.11	-0.08	-0.04	-0.01
Distance from Coast by <i>Fraxinus excelsior</i>	0.21	0.14	0.18	0.25	0.29
Distance from Coast by <i>Quercus robur</i>	-0.19	-0.25	-0.21	-0.16	-0.13
Elevation by <i>Alnus glutinosa</i>	0.07	0.02	0.05	0.08	0.11
Elevation by <i>Betula pendula</i>	0.02	-0.02	0.01	0.04	0.06
Elevation by <i>Fagus sylvatica</i>	-0.05	-0.10	-0.07	-0.03	0.01
Elevation by <i>Fraxinus excelsior</i>	-0.21	-0.30	-0.25	-0.18	-0.12
Elevation by <i>Quercus robur</i>	0.06	-0.00	0.03	0.08	0.12
Space Parameter by <i>Alnus glutinosa</i>	0.02	-0.01	0.01	0.03	0.05
Space Parameter by <i>Betula pendula</i>	-0.01	-0.04	-0.02	0.00	0.02

Table S6: Summary of Bernoulli model of false spring risk with varying rates of leafout for each species without the species interactions (estimates presented on logit scale for *Aesculus hippocastanum*).

	Estimate	10%	25%	75%	90%
Space Parameter by <i>Fagus sylvatica</i>	-0.09	-0.13	-0.11	-0.08	-0.05
Space Parameter by <i>Fraxinus excelsior</i>	-0.01	-0.07	-0.04	0.01	0.05
Space Parameter by <i>Quercus robur</i>	-0.00	-0.05	-0.02	0.01	0.04
Climate Change by <i>Alnus glutinosa</i>	0.06	0.03	0.05	0.08	0.10
Climate Change by <i>Betula pendula</i>	0.07	0.04	0.06	0.08	0.10
Climate Change by <i>Fagus sylvatica</i>	-0.47	-0.52	-0.49	-0.45	-0.43
Climate Change by <i>Fraxinus excelsior</i>	-0.49	-0.55	-0.51	-0.46	-0.42
Climate Change by <i>Quercus robur</i>	-0.42	-0.47	-0.45	-0.40	-0.38
NAO Index by Climate Change	-0.64	-0.68	-0.65	-0.62	-0.59
Mean Spring Temperature by Climate Change	0.59	0.54	0.57	0.61	0.64
Distance from Coast by Climate Change	-0.17	-0.22	-0.19	-0.14	-0.11
Elevation by Climate Change	0.13	0.08	0.11	0.15	0.19
Space Parameter by Climate Change	-0.07	-0.11	-0.09	-0.06	-0.04

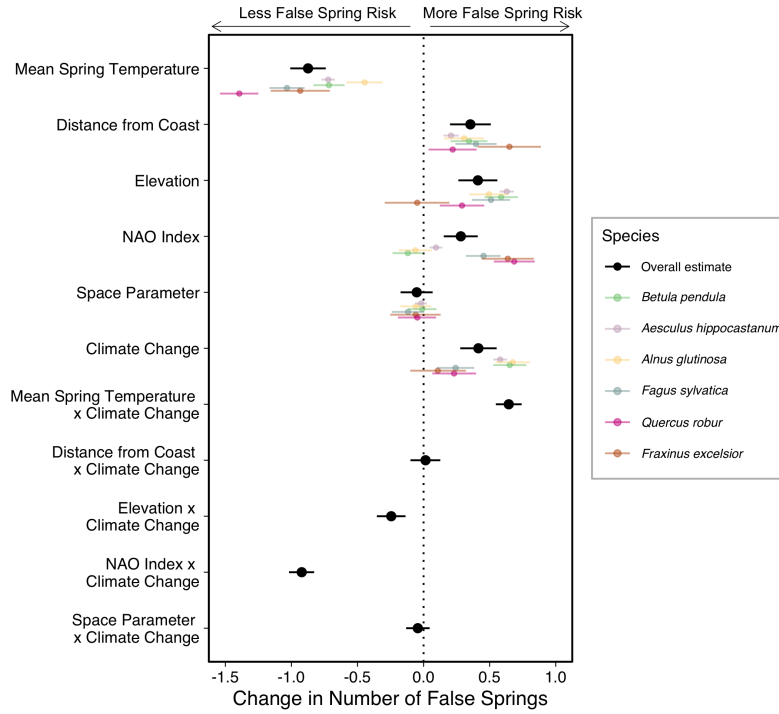


Figure S4: Effects of species, climatic and geographical predictors on false spring risk with a lower temperature threshold (-5°C) for defining a false spring. More positive values indicate an increased probability of a false spring whereas more negative values suggest a lower probability of a false spring. Dots and lines show means and 90% uncertainty intervals. Values closer to zero have less of an effect on false springs. There were 730,996 zeros and 23,855 ones for false springs in the data.

Table S7: Summary of Bernoulli model of false spring risk with a lower temperature threshold (-5°C) for defining a false spring without the species interactions (estimates presented on logit scale for *Aesculus hippocastanum*).

	Estimate	10%	25%	75%	90%
Intercept	-3.35	-3.37	-3.36	-3.34	-3.32
NAO Index	0.09	0.05	0.07	0.11	0.14
Mean Spring Temperature	-0.72	-0.77	-0.74	-0.70	-0.67

Table S7: Summary of Bernouilli model of false spring risk with a lower temperature threshold (-5°C) for defining a false spring without the species interactions (estimates presented on logit scale for *Aesculus hippocastanum*).

	Estimate	10%	25%	75%	90%
Distance from Coast	0.21	0.15	0.18	0.23	0.27
Elevation	0.63	0.58	0.61	0.65	0.68
Space Parameter	-0.02	-0.06	-0.04	0.00	0.03
Climate Change	0.58	0.53	0.56	0.60	0.63
Alnus glutinosa	0.33	0.28	0.31	0.34	0.37
Betula pendula	0.08	0.04	0.06	0.09	0.12
Fagus sylvatica	-0.45	-0.49	-0.46	-0.43	-0.40
Fraxinus excelsior	-1.58	-1.66	-1.61	-1.55	-1.50
Quercus robur	-1.26	-1.32	-1.28	-1.24	-1.20
NAO Index by Alnus glutinosa	-0.16	-0.23	-0.19	-0.12	-0.08
NAO Index by Betula pendula	-0.21	-0.28	-0.24	-0.19	-0.15
NAO Index by Fagus sylvatica	0.36	0.28	0.33	0.39	0.44
NAO Index by Fraxinus excelsior	0.54	0.39	0.48	0.60	0.69
NAO Index by Quercus robur	0.59	0.49	0.55	0.63	0.70
Mean Spring Temperature by Alnus glutinosa	0.28	0.19	0.24	0.31	0.36
Mean Spring Temperature by Betula pendula	0.01	-0.06	-0.02	0.03	0.07
Mean Spring Temperature by Fagus sylvatica	-0.31	-0.39	-0.35	-0.28	-0.23
Mean Spring Temperature by Fraxinus excelsior	-0.21	-0.39	-0.28	-0.14	-0.04
Mean Spring Temperature by Quercus robur	-0.67	-0.77	-0.71	-0.63	-0.58
Distance from Coast by Alnus glutinosa	0.10	0.01	0.06	0.14	0.19
Distance from Coast by Betula pendula	0.14	0.06	0.10	0.17	0.22
Distance from Coast by Fagus sylvatica	0.19	0.09	0.15	0.23	0.29
Distance from Coast by Fraxinus excelsior	0.44	0.26	0.37	0.52	0.62
Distance from Coast by Quercus robur	0.01	-0.11	-0.04	0.06	0.14
Elevation by Alnus glutinosa	-0.13	-0.23	-0.17	-0.10	-0.04
Elevation by Betula pendula	-0.04	-0.11	-0.07	-0.01	0.03
Elevation by Fagus sylvatica	-0.12	-0.21	-0.16	-0.08	-0.03
Elevation by Fraxinus excelsior	-0.68	-0.87	-0.75	-0.60	-0.49
Elevation by Quercus robur	-0.34	-0.45	-0.39	-0.29	-0.22
Space Parameter by Alnus glutinosa	-0.04	-0.11	-0.07	-0.01	0.03

Table S7: Summary of Bernouilli model of false spring risk with a lower temperature threshold (-5°C) for defining a false spring without the species interactions (estimates presented on logit scale for *Aesculus hippocastanum*).

	Estimate	10%	25%	75%	90%
Space Parameter by <i>Betula pendula</i>	0.01	-0.05	-0.02	0.03	0.07
Space Parameter by <i>Fagus sylvatica</i>	-0.10	-0.18	-0.13	-0.07	-0.02
Space Parameter by <i>Fraxinus excelsior</i>	-0.04	-0.19	-0.10	0.02	0.10
Space Parameter by <i>Quercus robur</i>	-0.03	-0.13	-0.07	0.01	0.07
Climate Change by <i>Alnus glutinosa</i>	0.10	0.02	0.06	0.13	0.17
Climate Change by <i>Betula pendula</i>	0.07	-0.00	0.04	0.10	0.14
Climate Change by <i>Fagus sylvatica</i>	-0.34	-0.43	-0.37	-0.30	-0.25
Climate Change by <i>Fraxinus excelsior</i>	-0.47	-0.63	-0.54	-0.41	-0.31
Climate Change by <i>Quercus robur</i>	-0.35	-0.46	-0.40	-0.30	-0.23
NAO Index by Climate Change	-0.92	-1.02	-0.96	-0.88	-0.83
Mean Spring Temperature by Climate Change	0.64	0.55	0.60	0.69	0.74
Distance from Coast by Climate Change	0.01	-0.10	-0.03	0.06	0.13
Elevation by Climate Change	-0.24	-0.35	-0.29	-0.20	-0.14
Space Parameter by Climate Change	-0.04	-0.13	-0.08	-0.01	0.05