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# ▶ To cite this version:

Ervin Rasztovits, Imre Berki, Csaba Mátyás, Kornél Czimber, Elisabeth Pötzelsberger, et al.. The incorporation of extreme drought events improves models for beech persistence at its distribution limit. Annals of Forest Science, 2014, 71 (2), pp.201 - 210. 10.1007/s13595-013-0346-0. hal-01098420

# HAL Id: hal-01098420 https://hal.science/hal-01098420

Submitted on 24 Dec 2014  $\,$ 

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# ORIGINAL PAPER

# The incorporation of extreme drought events improves models for beech persistence at its distribution limit

Ervin Rasztovits • Imre Berki • Csaba Mátyás • Kornél Czimber • Elisabeth Pötzelsberger • Norbert Móricz

Received: 3 August 2012 / Accepted: 15 November 2013 / Published online: 29 November 2013 © INRA and Springer-Verlag France 2013

#### Abstract

• *Context* Projections of species distribution models under future climate are usually based on long-term averages. However, singular extreme drought events presumably contribute to the shaping of distribution limits at the retreating low-elevation xeric limits.

• *Methods* The objectives of this study were to set up a distribution model based on extreme drought events (EDM), which uses sanitary logging information as a proxy of vitality

# Handling Editor: Erwin Dreyer

**Contribution of co-authors** Ervin Rasztovits: writing the paper and analysing the data. Norbert Móricz: writing the paper and analysing the data.

Imre Berki: supervising the work. Elisabeth Pötzelsberger: climate data analyses.

Kornél Czimber: GIS analyses.

Csaba Mátyás: supervising the work and coordinating the research project.

**Electronic supplementary material** The online version of this article (doi:10.1007/s13595-013-0346-0) contains supplementary material, which is available to authorized users.

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Energy Department, Austrian Institute of Technology, Donau-City-Straße, 1220 Vienna, Austria response of beech, and compare it with the results of classical species distribution models (SDMs).

• *Results* Predictions of the EDM for 2025 were in agreement with those of the SDM, but EDM predicted a more serious decline in all regions of Hungary towards the end of the century.

• *Conclusion* These results suggest that the predicted increase in frequency and severity of drought events may further limit the distribution of beech in the future.

**Keywords** Beech · Trailing edge · Climate change · Xeric limit · Predictive modelling

# **1** Introduction

European beech is a strong competitor with high shadow tolerance and able to dominate on sites with a relatively broad spectrum of nutrient and hydrological regimes (Leuschner et al. 2006). In the southern part of its range at low-elevation xeric limits (Mátyás et al. 2009), beech is limited by increasing water stress, and it is outcompeted by more drought-tolerant species, mainly oaks and pines (Horvat et al. 1974; Ellenberg 1986). Drought sensitivity is assumed to be a key factor limiting growth and distribution of beech near to its lower distribution limit (xeric limit; Czúcz et al. 2010) in southern and south-eastern Europe.

Several studies suggest a decline in beech regeneration (Peñuelas et al. 2007; Betsch et al. 2011; Aranda et al. 2012), rapid growth decline (Jump et al. 2006; Piovesan et al. 2008) and extensive range retraction (Czúcz et al. 2010; Kramer et al. 2010) with warmer and drier climatic conditions. Consequently, modelling the vitality response of beech to predicted changes of climate is a critical issue (Franke and Köstner 2007).



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Species distribution models (SDMs) are extensively used in ecology and management planning but first of all for predicting species distribution under future climates. SDMs connect the species' environmental requirements of the localities where it is currently known to occur without specific ecological knowledge (Guisan and Zimmermann 2000). They can be evaluated for their ability to predict current distributions, but it is not tested whether models that are successful in predicting current distributions are equally powerful in predicting distributions under future climates. Studies comparing modelling algorithms are now common (Tsoar et al. 2007), but Thuiller et al. (2004) have pointed out the problem of large variation among SDM predictions.

The multidimensional climatic envelope created by niche-based models is determined by long-term averages. However, long-term climatic means do not express the effect of extreme weather events, which trigger vitality decline, appearance of pests or diseases leading to mortality (Löw et al. 2006, Geßler et al. 2007; Lakatos and Molnár 2009). The probability of drought events is expected to increase in the second half of the twenty-first century (Schär et al. 2004). Furthermore, the number of consecutive dry years and dry summers may also increase. Southeast Europe is one of the climatically most exposed regions on the continent although largely neglected by studies (Mátyás et al. 2010). For the period 2051–2100, the severity of drought events may increase significantly in all scenarios compared with the control period 1951-2000 (Gálos et al. 2007).

A significant drought event occurred between 2000 and 2003 in South-West of Hungary, which was unprecedented in duration and strength since the beginning of the 1950s. After this drought event, large volume of dead or already declining beech was logged by forest managers under the control of the Forest Directorates of the Central Agricultural Office. First, solitary trees showed the typical symptoms of reduced water availability (leaf yellowing, top drying) in 2002. The symptoms of xylo- and phloeophagous insect attack (*Agrilus viridis*, *Taphrorychus bicolor*) and fungal infection (*Biscogniauxia nummularia*, *Nectria coccinea*) appeared in 2003 and expanded rapidly after 2004 (Lakatos and Molnár 2009).

SDMs assume that the modelled species is in equilibrium with its environment. Although this is a required assumption for projecting the model in space, a few critical considerations have been raised in the recent literature on how close a system really is to an equilibrium and how long it would take to reach a new equilibrium, e.g. after an environmental change. Svenning and Skov (2004) measured low range filling for many European tree species, suggesting that many of these species might not be in equilibrium with their

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environment throughout their whole range. The nonequilibrium consideration is a critical issue in modelling the distribution of invasive or retreating species.

In this study, the regional effects of a singular drought event were used to develop an extreme drought event-based model (EDM), which captures in space and time the specific momentum, when the modelled system is tipped out from its equilibrium state. We evaluated the outputs of SDMs and compared the predictions of the best performing SDM with the results of our EDM. Modelling was concentrated to Hungary where the retreat of beech is expected to be imminent.

We addressed the following questions:

- 1. Which SDM can best describe the present distribution of beech at its xeric limit in Hungary?
- 2. How can the effects of a (singular) drought event be modelled?
- 3. What are the differences in the projections for the potential future distribution of beech using SDMs and the EDM?

To answer the research questions, we performed the following: (1) modelled the current and potential future distribution of beech in Hungary using SDMs; (2) to obtain vitality response of beech, we related sanitary logging information as a proxy of vitality condition to meteorological data; and (3) modelled and compared predictions for the future vitality status of beech in the course of this century, applying a regional climate model projection.

# 2 Material and methods

# 2.1 SDMs using long-term climate data

# 2.1.1 Environmental layers and variable selection

Environmental variables included climate, soil and geomorphological data, all with a spatial resolution of  $1 \times 1$  km. Climate data for the past were obtained from the Hungarian Meteorological Service, while future climate data were derived from the World Climate Center Database. The soil data originated from the Centre for Agricultural Research (Hungarian Academy of Sciences) and topographical data from the SRTM digital elevation model.

For the description of climate, various monthly temperature and precipitation data plus a set of 19 climatederived variables were used (48 variables in total). In addition, we used 12 bioclimatic indices (23 variables in total) computed from minimum and maximum monthly averaged temperatures and monthly precipitations (Online resource 1). For the past period (1950–2000), air temperature and precipitation maps were interpolated by kriging, using data of 31 stations in case of air temperature and approx. 600 rain gauges in case of precipitation. We used for the elevation dependence of temperature monthly vertical gradients (Péczely 1979). The effect of slope and aspect on air temperature was considered by assuming a linear relationship between air temperature and global radiation (Xin et al. 2007). The reliability of the climate maps was assessed by cross-validation.

For the future climate simulations (2000–2100), we applied the 'Climate Limited area-Modelling' (CLM) regional climate model using the A1B scenario (mean of two runs) with a grid size of 0.2° (Keuler et al. 2009). CLM model data were corrected using the delta change approach (Hay et al. 2000), based on the mean deviation of the observed and simulated variables between 1960 and 2000 for each grid box. Corrected long-term averages of monthly air temperature and precipitation were derived for the periods 2011–2040, 2036–2065 and 2066–2095.

Three soil variables (soil texture, soil moisture regime and genetic soil type; AGROTOPO 2002) and further three topographical factors (mean altitude, slope and aspect) were also added.

To determine which variables to exclude, redundant environmental layers were identified via pairwise correlation. Five thousand randomly selected points were used to calculate the Pearson correlation. Variables showing a correlation >0.80 were considered redundant. Between any two redundant variables, those related to climate extremes were preferred.

# 2.1.2 Occurrence data

Data for the habitat suitability modelling were derived from the Hungarian Forest Inventory (NFI) database, actualised for 2010, including all the forest subcompartments containing beech (above the 5 % threshold limit).

In total, 11,332 subcompartments were considered as 'true presence' observation points. For the presence–absence methods, the same number of 'pseudo-absence' points was created randomly throughout the country with a buffer zone of 1,000 m around the presence points.

# 2.1.3 Modelling algorithms

The ModEco (Guo and Liu 2010) software was chosen for the SDMs, since it includes tools for feature analysis, model performance evaluation and an accuracy assessment tool, supporting consistent comparisons. We assessed and compared three presence–absence methods: the generalised linear model (GLM), the artificial neural network using back-propagation algorithm (BP-ANN, Maravelias et al. 2003) and the classification tree (CTree, Breiman et al. 1984).

# 2.1.4 Accuracy assessment

The cross-validation test was applied and evaluated. Data were randomly split into 10 subsets of equal size; each subset was in turn used for accuracy testing and the remaining nine subsets for training. Finally, the total accuracy was estimated after an overlaying operation by averaging the accuracy of each test. Maximum kappa value and the true positive rate (TPR: the proportion of the positive hits in the prediction) were used to assess model performance (Elith et al. 2006).

# 2.2 Extreme drought event-based model

# 2.2.1 Study area

Beech is one of the dominant tree species in Southern Hungary, Zala County, with a share of 17.4 % of the forest cover (Fig. 1). Preference is given to the species due to high growth and economic value.

The climate of the region is mild and relatively humid, and beech is considered to be in its optimum there. Climatic conditions, especially precipitation, show a strong Northeast–Southwest gradient (mean annual precipitation sum from 650 to 800 mm) increasing towards the Slovenian and Croatian state border. Mean annual temperature varies between 9 and 10.5 °C, where the mean temperature of the warmest month is 19.2 °C and the mean of the coldest is -1.5 °C.

The elevation of the hilly landscape varies between 120 and 380 m a.s.l., and 88 % of beech forests are situated on deep loam soil (Halász 2006).

# 2.2.2 Identification of the drought period

The annual aridity index (Budyko 1974) was calculated for three regional meteorological stations between 1951 and 2010 to characterise the climate trend in the study area. Chow's *F* statistics (Chow 1960) were computed for potential significant (at 97.5 % confidence limit) trend breaks using the '*R*' package 'strucchange' (Zeileis et al. 2002). The year 2000 was identified as first breakpoint and 2004 as the second trend breakpoint at all meteorological stations. Consequently, the drought between 2000 and 2003 was considered as a singular extreme event, significantly different from the long-term trend. Chow's *F* statistics for the meteorological station Szentgotthárd–Farkasfa is presented in the Online Resource 2.



Fig. 1 Distribution of beech in Hungary based on Forest Inventory data 2011 (study area is highlighted in *red*; the meteorological station Szentgotthárd–Farkasfa is indicated)



# 2.2.3 Sanitary logging data

Symptoms of vitality loss and mortality triggered by drought events are extremely difficult to trace in a larger area due to their spontaneous and disperse character. As a proxy, annual felling volume data of sanitary logging were gathered in the two involved State Forest Companies (Szombathelyi Erdőgazdaság Zrt., Zalaerdő Zrt) for each subcompartment for the period 2000–2008. Sanitary logging affected 14.3 % of the beech forest subcompartments with a total area of 4189 ha. The operations were supervised by forest inspectors, so data are relatively reliable, except for some inaccessible stands left unlogged.

# 2.2.4 Defining the vitality response of beech

The modified Ellenberg climate quotient  $(EQ_m)$  was chosen as predictor for the vitality response of beech, since this was the highest ranked index, which simultaneously considered the effects of precipitation and temperature (Ellenberg 1986). The interpolated meteorological surfaces were used for computing the 4 year mean (2000–2003) of the EQ<sub>m</sub>, based on the results of the drought delineation.

#### 2.2.5 Climate model data

The CLM regional climate model was applied also for the EDM. Long-term precipitation and air temperature differences were computed between the observed mean of the grid boxes and each beech subcompartment for the period 1960–2000. The differences were added to the mean-altitude simulation values corrected previously by the delta change approach. The computed values were subsequently used for calculating the annual  $EQ_m$ 's for each subcompartment from 2011 to 2100.

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#### 2.2.6 Simulation of the future vitality condition of beech

The relationship between mean  $EQ_m$  and volume of sanitary logging was analysed to obtain the vitality response of beech. Sanitary logging data between 2000 and 2008 were pooled, since fellings could not always keep up with the decay.

 $EQ_m$  values of all beech subcompartments based on the  $EQ_m$  were divided into 18 equal intervals interpreted as drought classes. For every drought class, the total area occupied by beech was calculated. The affected area per class was expressed in presence of the total area.

The defined vitality response was used to simulate the future conditions until 2025, 2050 and 2100, based on the mean  $EQ_m$ 's of the worst four subsequent years.

# **3 Results**

3.1 SDMs based on long-term climate data

#### 3.1.1 Capturing of the current potential distribution

GLM performed relatively poorly, which resulted from strong overprediction. CTree and BP-ANN methods performed notably better. High TPR, smaller predicted potential area and high kappa score indicated that these models are able to capture non-linear responses and can manage interactions between the variables (Table 1).

#### 3.1.2 Prediction of future distribution

GLM overpredicted the distribution of beech in the near future; it included some regions already outside of the current distribution range. BP-ANN predicted almost no reduction in the potential area for the period 2011–2040 and a very slight reduction (8 %) for 2036–2065. A considerable shrinking (57 %) was predicted only by the end of this century, indicating that 45 % of the current

 Table 1
 Parameters and statistical

 performance of presence/absence
 models

Model	Parameters	True positive rate (TPR)	Predicted area rate	Kappa index
Artificial neural network with back-propagation (BP-ANN)	Momentum 0.3 Learning rate 0.1	0.9425	1.2096	0.8336
Classification tree (CTree)	Number of trails 10 Window size 20	0.9493	1.3196	0.8431
	Pruning conf. level 0.25			
Generalised linear model (GLM)	Link function type LOGIT Threshold 0.426	0.9592	1.6237	0.8174

stands may be outside of the potential area. Regionally, the most serious decrease was predicted for the lowelevation hills in the Southwest. CTree predicted a more pronounced shrinkage in all regions of Hungary by losing 37, 68 and 75 %, respectively (Fig. 2).

# 3.1.3 Factor importance analysis

Factor importance analysis is algorithm sensitive, but among the environmental variables, the maximum temperature of summer months and the  $EQ_m$  appeared repeatedly as the most influential predictor (Table 2).

# 3.2 Extreme drought event-based model

# 3.2.1 Vitality response of beech

With increasing  $EQ_m$  values, beech vitality is expected to decline, leading to partial or total tree mortality (Fig. 3).

Due to the observed irregular response to increasing drought, application of a continuous function was rejected. Instead, we divided the range of distribution into three classes: (1) EQ<sub>m</sub> less than 53 with no damage, (2) EQ<sub>m</sub> from 53 to 65 with moderate damage (mean damage ratio 12.7 %) and (3) EQ<sub>m</sub> above 65 with serious damage (mean damage ratio 55.4 %). The 95 % confidence intervals were separated quite well at the boundary of the moderate and serious damage classes.

# 3.2.2 Simulation results of beech vitality

Vitality decline of beech is predicted only after 2025; until 2025, extreme drought events might cause only local damages along the xeric distribution limit (Fig. 4a–c).

By 2050, serious vitality decline is expected regionally not only at the lower distribution limit but also at sites considered earlier as optimal. Moderate damage will be likely at almost all beech sites, except mountainous regions, approximately above 500–600 m. By end of the century, beech might be seriously affected by extreme events in its entire Hungarian range, except above 700–800 m in the Northeast.

# 4 Discussion

# 4.1 Performance of the SDMs

Mathematical properties of the SDMs can help to explain the differences in their predictive performance. GLM performed relatively poorly due to the lack of flexibility (Austin 2002). CTree provided among all models the best statistical performance for the current distribution, although the predictions for the future showed regional inconsistency especially in the Southwest and in the Northeast. The relatively good predictive performance of CTree could be explained by the ability to find interactions and hierarchical relations among environmental variables (Austin 2002). Although BP-ANN performed slightly poorer than CTree, its predictions for the future were more realistic without regional inconsistency. ANNs have been already used with great success in a variety of species habitat/bioclimatic suitability analyses (Araújo et al. 2005). The strength of ANN is the ability to 'learn' underlying (nonlinear) patterns of correlation between observed input (environmental/climatic variables) and target (species presence/ absence) data.

All models predicted a larger potential distribution area for the present than the current one. The difference between the potential and realised range could be explained mainly by two factors: human interference land use change and the lack of more detailed soil information. The occurrence of beech is often limited by unfavourable textural and hydrological soil characteristics (pseudogley).

# 4.2 Performance of the EDM

The main advantage of the extreme drought event-based approach is that it uses the empirical relationship between observed drought severity and damage. Since our



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Fig. 2 Potential distribution modelled by BP-ANN and CTree for present and future conditions (2011–2040, 2036–2065 and 2066–2095), respectively. *Green colour* represents areas modelled as suitable for beech at the given period

model applies sanitary logging data, it provides primarily silvicultural rather than ecological substration. It is important to emphasise that serious damage indicated by EDM does not mean implicitly extinction. Moreover, the aim of sanitary logging is first of all to salvage valuable timber from the damaged compartments; thus, healthy trees are also often logged. Past events have already shown that decline of forests is triggered by extreme drought periods (Zimmermann et al. 2009). Symptoms of decline appear usually rather abruptly. Our results suggest that drought-induced mortality may be connected to a certain threshold and large-scale decline may develop rapidly during and following droughts. In addition to biotic factors such as pathogens and insects, two physiological

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 Table 2
 The overall classification accuracy of the models and the most predictive five factors with the related kappa values that resulted from the factor importance analyses

Rank	Models					
	CTree		GLM			
	Predictor	Kappa	Predictor	Kappa		
	Overall	0.843	Overall	0.817		
1	Tmax_05	0.717	Tmax_05	0.708		
2	Tmax_06	0.707	Tmax_06	0.697		
3	Tmax_08	0.704	Tmax_07	0.673		
4	Tmax_04	0.704	EQm	0.670		
5	EQm	0.673	Tmean_05	0.664		

mechanisms have been proposed in the literature as potential causes of mortality: (1) embolism happens rapidly when soil water potential drops below a threshold value and if there is no efficient stomatal limitation of water losses (Lemoine et al. 2002), and (2) the C starvation hypothesis predicts that reduced C assimilation caused by drought-induced stomatal closure leads to an imbalance between C availability and C loss. Over time, if drought persists, such negative C balance can lead to an exhaustion of C reserves and, ultimately, to death. As beech trees have large stored C pools regardless of past stress (Sala et al. 2010), the hydraulic failure is the more probable reason for drought-induced mortality. However, hydraulic failure is difficult to trace in the field. Lower resistance to cavitation (Rice et al. 2004) and lethal leaf dehydration (Kursar et al. 2009) have been related to increased mortality in other species, rendering hydraulic failure a very plausible mechanism for observed tree mortality under severe, acute drought.

A drawback of the EDM is that well-documented system tip-outs are sparse in space and time; therefore, such models are based on local data. The spatial extension of the locally

Fig. 3 Relationship of the modified Ellenberg climate quotient and percentage of area affected intervention by sanitary with 95 % confidence intervals (explanation in the text)

observed response to the whole range in case of a widespread species needs special attention.

Simulation results using the EDM have further uncertainties. The stochastic projection of drought may influence considerably the results at a specific location. Even minor changes of precipitation trends influence considerably the future vitality of beech, especially close to the xeric limit.

The spatial uncertainty and temporal randomness of drought occurrence was reduced by using the mean of two runs of the A1B scenario and by using long periods for projections. Soil physical properties did not appear as key environmental factors due to their low informative value. Soil water holding capacity is considered as critical during severe drought periods (Gärtner et al. 2008), and its better assessment should contribute to the certainty of the predictions.

Biotic uncertainty results partly from the limited ability to predict the future role of consumers and pathogens; previously unknown pests may appear, with unpredictable consequences. Further, the potential persistence of forest trees is another source of uncertainty, e.g. the role of phenotypic plasticity (Mátyás 2007). The persistence of forests is further supported by planned forest management, which may assist to maintain beech forests in the future (Mátyás et al. 2009).

#### 4.3 Comparisons of predictions

The best SDM model displayed considerable regional differences, and EDM predicted more severe dieback for the middle and end of this century.

There was almost no difference between the two model predictions for 2025. BP-ANN predicted no reduction in the potential area while the EDM predicted serious damage only insignificantly and medium damage on 23.1 % of the range mainly close to the margins.





Fig. 4 Beech vitality simulation by the extreme drought-based model (EDM) for 2025 (a), 2050 (b) and 2100 (c) in Hungary using the A1B scenario of the CLM climate model (*dark green* healthy stands, *yellow* moderate damage, *red* serious decline)



The difference between the two approaches gets visible only in 2050, where BP-ANN predicts only minor shrinking of the potential range (15.0 %), while EDM estimates damages on 84.3 % of the range.

By the end of this century, EDM predicts stability problems practically on the whole range (99.9 %) while BP-ANN predicts a reduction only to 43.2 %. The most serious decrease is predicted for the Southwest using BP-ANN, while EDM produces a spatially more homogeneous and more pronounced vitality loss. The main causes of the differences can be attributed to the fact that widespread mortality is caused by extreme events and not by the shift of climate means.

# **5** Conclusion

The three investigated SDMs performed fair in describing the current distribution of beech, but machine learning methods like CTree and BP-ANN have shown better results.  $EQ_m$  has been chosen as main environmental predictor in the EDM.

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Coupling sanitary logging information with  $EQ_m$  produced an abrupt decline of the vitality at a certain level of drought conditions. Prediction for 2025 obtained from the EDM was in agreement with those of the SDM, but to the end of the century, the EDM predicted a more serious decline in all regions of Hungary. The result of the comparison suggests that in the future, the increasing frequency and severity of extremes will play a more important role in limiting the distribution of beech (and most probably also of other species) than long-term climate means.

**Acknowledgments** We would like to acknowledge Dr Tibor Szép who helped in providing the occurrence and sanitary logging data. We also thank Prof. Dr Hubert Hasenauer for his personal communication regarding the methods of modelling.

**Funding** This research was funded by the Austrian–Hungarian Transboundary Cooperation 2007–2013 ('FaKlim' project—L00044), by TÁMOP-4.2.2.A-11/1/KONV-2012–0013 and by the FORGER ('Towards the Sustainable Management of Forest Genetic Resources in Europe'—289119) project.

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