

1 **Legacies of past land use have a stronger effect on forest carbon exchange**
2 **than future climate change in a temperate forest landscape**

3

4 **Running head:** “Land use legacies determine C exchange”

5

6 Dominik Thom*^{1,2}, Werner Rammer¹, Rita Garstenauer^{3,4}, Rupert Seidl¹

7 ¹ Institute of Silviculture, Department of Forest- and Soil Sciences, University of Natural
8 Resources and Life Sciences (BOKU) Vienna, Peter-Jordan-Straße 82, 1190 Vienna, Austria

9 ²Rubenstein School of Environment and Natural Resources, University of Vermont, 308i Aiken
10 Center, Burlington, VT 05405, USA. Tel: +1 802 557 8221. Fax: +1 802 656 2623. Email:
11 dominik.thom@uvm.edu

12 ³ Institute of Social Ecology, Department of Economics and Social Sciences, University of
13 Natural Resources and Life Sciences (BOKU) Vienna, Peter-Jordan-Straße 82, 1190 Vienna,
14 Austria

15

16 * Corresponding author

17 **Abstract**

18 Forest ecosystems play an important role in the global climate system, and are thus intensively
19 discussed in the context of climate change mitigation. Over the past decades temperate forests
20 were a carbon (C) sink to the atmosphere. However, it remains unclear to which degree this C
21 uptake is driven by a recovery from past land use and natural disturbances or ongoing climate
22 change, inducing high uncertainty regarding the future temperate forest C sink. Here our
23 objectives were (i) to investigate legacies within the natural disturbance regime by empirically
24 analyzing two disturbance episodes affecting the same landscape 90 years apart, and (ii) to
25 unravel the effects of past land use and natural disturbances as well as future climate on 21st
26 century forest C uptake by means of simulation modelling. We collected historical data from
27 archives to reconstruct the vegetation and disturbance history of a forest landscape in the
28 Austrian Alps from 1905 to 2013. The effects of legacies and climate were disentangled by
29 individually controlling for past land use, natural disturbances, and future scenarios of climate
30 change in a factorial simulation study. We found only moderate spatial overlap between two
31 episodes of wind and bark beetle disturbance affecting the landscape in the early 20th and 21st
32 century, respectively. Our simulations revealed a high uncertainty about the relationship
33 between the two disturbance episodes, whereas past land use clearly increased the impact of the
34 second disturbance episode on the landscape. The future forest C sink was strongly driven by
35 the cessation of historic land use, while climate change reduced forest C uptake. Compared to
36 land use change the two past episodes of natural disturbance had only marginal effects on the
37 future carbon cycle. We conclude that neglecting legacies can substantially bias assessments of
38 future forest dynamics.

39

40 **Key words:** bark beetles, climate change, forest history, forest management, Kalkalpen
41 National Park, legacy effects, net ecosystem exchange, wind

42

43

44 **Copyright statement**

45 The authors agree to the copyright statement as described at
46 https://www.biogeosciences.net/about/licence_and_copyright.html.

47

48 **1. Introduction**

49 Carbon dioxide (CO₂) is responsible for 76% of the global greenhouse gas emissions, and is
50 thus the single most important driver of anthropogenic climate change (IPCC 2014). Forest
51 ecosystems take up large quantities of CO₂ from the atmosphere, and play a key role in
52 mitigating climate change (IPCC 2007). During the period 1990 – 2007, established and
53 regrowing forests were estimated to have taken up 60% of the cumulative fossil carbon
54 emissions (Pan et al., 2011). This carbon (C) sink strength of forests has further increased in
55 recent years (Keenan et al., 2016), resulting from multiple drivers: On the one hand, possible
56 factors contributing to an increasing sink strength of the biosphere are CO₂ (Drake et al., 2011)
57 and nitrogen (Perring et al., 2008) fertilization, in combination with extended vegetation periods
58 resulting from climate warming (Keenan et al., 2014). On the other hand, the accelerated carbon
59 uptake of forests might be a transient recovery effect of past carbon losses from land use and
60 natural disturbances (Erb, 2004; Loudermilk et al., 2013).

61 For the future, dynamic Global Vegetation Models (DGVMs) frequently suggest a persistent
62 forest carbon sink (Keenan et al., 2016; Sitch et al., 2008). However, while DGVMs are suitable

63 for tracking the direct effects of global change, they frequently neglect the effects of long-term
64 legacies of the past. Both natural disturbances (e.g., wind storms and bark beetle outbreaks) and
65 land use have decreased the amount of carbon currently stored in forest ecosystems (Erb et al.,
66 2018; Goetz et al., 2012; Harmon et al., 1990; Seidl et al., 2014a). The legacy effects of past
67 disturbances and land use have the potential to significantly influence forest dynamics and alter
68 the trajectories of carbon uptake in forest ecosystems over time frames of decades and centuries
69 (Gough et al., 2007; Landry et al., 2016; Seidl et al., 2014b). This is of particular importance
70 for the forests of Central Europe, which have been markedly affected by forest management
71 and natural disturbances over the past centuries (Naudts et al., 2016; Svoboda et al., 2012). The
72 importance of an improved understanding of past disturbance dynamics and its impacts on the
73 future carbon cycle is further underlined by the expectation that climate change will amplify
74 natural disturbance regimes in the future (Seidl et al., 2017). In this context the role of temporal
75 autocorrelation within disturbance regimes is of particular relevance, i.e., the influence that past
76 disturbances and land use have on future disturbances at a given site. Are past disturbances and
77 land use increasing or decreasing the propensity and severity for future disturbances? And are
78 such temporal autocorrelations influencing the future potential of forests to take up carbon? The
79 propensity and effect of such interactions between disturbances and land use across decades
80 remain understudied to date, largely due to a lack of long-term data on past disturbances and
81 land use.

82 Here we investigate the effect of long-term disturbance and land use legacies on forest
83 ecosystem dynamics, in order to better understand the drivers of future forest carbon uptake,
84 and thus aid the development of effective climate change mitigation strategies. In particular,
85 our first objective was to investigate the temporal interaction of two major episodes of natural
86 disturbance affecting the same Central European forest landscape 90 years apart (i.e., 1917 –
87 1923 and 2007 – 2013). We hypothesized a temporal autocorrelation of the two major

88 disturbance episodes, and specifically an amplifying effect from the earlier disturbance episode
89 on the later disturbance episode(see e.g., Schurman et al., 2018). Our hypothesis was based on
90 the importance of landscape topography for wind and bark beetle disturbances (Senf and Seidl,
91 2018; Thom et al., 2013), and the fact that susceptibility to these agents generally increases with
92 stand age, and is usually high after 90 years of stand development (Overbeck and Schmidt,
93 2012; Valinger and Fridman, 2011). In addition, we tested the effect of land use on the more
94 recent natural disturbance episode, following the hypothesis that land use increased natural
95 disturbance risk in Central Europe by promoting homogeneous structures and single-species
96 plantations (Seidl et al., 2011; Silva Pedro et al., 2015). Our second goal was to quantify the
97 contribution of past natural disturbance and land use on the future C uptake of the landscape
98 under a number of climate change scenarios using simulation modelling. We were particularly
99 interested in the relative effects of past disturbance, land use, and future climate on the future
100 forest C sink strength. To that end we reconstructed the vegetation history of the landscape from
101 1905 to 2013 using historical sources and remote sensing. We subsequently determined the
102 effect of past disturbance and land use on 21st century C dynamics by simulating forests from
103 the early 20th century to the end of the 21st century, experimentally altering past disturbance
104 and land use regimes in a factorial simulation experiment. These analyses were run under
105 multiple climate scenarios for the 21st century, and focused on Net Ecosystem Exchange (NEE)
106 (i.e., the net C exchange of the ecosystem with the atmosphere, which is the inverse of Net
107 Ecosystem Productivity, NEP) as the response variable. We hypothesized that the legacies of
108 past disturbance and land use are of paramount importance for the future carbon sink (Gough
109 et al., 2007; Thom et al., 2017a), expecting a saturation of carbon uptake as the landscape
110 recovers from past disturbance and land use (i.e., a negative but decreasing NEE through the
111 21st century). Moreover, we hypothesized a negative impact of future climate change on carbon
112 uptake as a result of less favorable conditions for carbon-rich spruce dominated forests
113 (Kruhlov et al., 2018; Thom et al., 2017a).

114

115 **2. Materials and Methods**

116 **2.1 Study area**

117 We selected a 7,609 ha forest landscape located in the northern front range of the Alps as our
118 study area (Fig. 1). Focusing on the landscape scale allowed us to mechanistically capture
119 changes in forest structure and C stocks by jointly considering large scale processes such as
120 disturbances as well as fine scale processes such as competition between individual trees. The
121 focal landscape is particularly suited to address our research questions as it (i) was affected by
122 two major episodes of natural disturbance (driven by wind and bark beetles) in the past century,
123 and (ii) has a varied land use history, with intensive management up until 1997, and then
124 becoming a part of Kalkalpen National Park (KANP), the largest contiguous protected forest
125 area in Austria. The steep elevational gradient of the study landscape, ranging from 414 m to
126 1637 m a.s.l., results in considerable variation in environmental conditions. For instance,
127 temperatures range from 4.3 – 9.0°C and mean annual precipitation sums vary between 1179 –
128 1648 mm across the landscape. Shallow Lithic and Renzic Leptosols as well as Chromic
129 Cambisols over calcareous bedrock are the prevailing soil types (Kobler 2004). The most
130 prominent natural forest types on the landscape are European beech (*Fagus sylvatica* [L.])
131 forests at low elevations, mixed forests of Norway spruce (*Picea abies* [K.]), silver fir (*Abies*
132 *alba* [Mill.]) and European beech at mid-elevations, and Norway spruce forests at high
133 elevations. These forest types are among the most common ones in Europe, and are highly
134 valuable to society also from a socio-economic perspective (Hanewinkel et al., 2012).

135

136 **2.2 Simulation model**

137 We employed the individual-based forest landscape and disturbance model (iLand) to simulate
138 past and future forest dynamics at our study landscape. iLand is a high-resolution process-based
139 forest model, designed to simulate the dynamic feedbacks between vegetation, climate,
140 management and disturbance regimes (Seidl et al., 2012a, 2012b). It simulates processes in a
141 hierarchical multi-scale framework, i.e., considering processes at the individual tree (e.g.,
142 growth, mortality as well as competition for light, water, and nutrients), stand (e.g., water and
143 nutrient availability), and landscape (e.g., seed dispersal, disturbances) scale as well as their
144 cross-scale interactions. Competition for resources among individual trees is based on
145 ecological field theory (Wu et al., 1985). Resource utilization is modelled employing a light use
146 efficiency approach (Landsberg and Waring, 1997), incorporating the effects of temperature,
147 solar radiation, vapor pressure deficit as well as soil water and nutrient availability on a daily
148 basis. Resource use efficiency is further modified by variation in the atmospheric CO₂
149 concentration. Seeds are dispersed via species-specific dispersal kernels (20 × 20 m horizontal
150 resolution) around individual mature trees. The establishment success of tree regeneration is
151 constrained by environmental filters (e.g., temperature and light availability). Mortality of trees
152 is driven by stress-induced carbon starvation and also considers a stochastic probability of tree
153 death depending on life-history traits.

154 Climate change affects tree growth and competition in iLand in several ways (Seidl et al.,
155 2012a, 2012b). For instance, an increase in temperature modifies leaf phenology and the length
156 of the vegetation period, but also reduces soil water availability due to increased
157 evapotranspiration. Net primary production is further influenced by climate change-induced
158 alterations in precipitation, atmospheric CO₂ levels, and solar radiation. Trees respond
159 differently to changes in climate in iLand based on their species-specific traits. Climate change
160 thus not only alters biogeochemical processes in the model but also modifies the competitive
161 strength of tree species, and consequently forest composition and structure (Thom et al. 2017a).

162 iLand currently includes three submodules to simulate natural disturbances, i.e., wind (Seidl et
163 al., 2014c), bark beetles (Seidl and Rammer 2017), and wildfire (Seidl et al., 2014b). As wind
164 and bark beetles are of paramount importance for the past and future disturbance regimes of
165 Central Europe's forests (Seidl et al., 2014a; Thom et al., 2013), we employed only these two
166 process-based disturbance submodules in our simulations. The impact of wind disturbance in
167 iLand depends on species- and size-specific susceptibility (e.g., critical wind speeds of
168 uprooting and stem breakage), vertical forest structure (e.g., gaps), and storm characteristics
169 (e.g., maximum wind speeds). The bark beetle module simulates the impact of *Ips typographus*
170 (L.) on Norway spruce, and thus addresses the effects of the most important bark beetle species
171 in Europe with respect to area affected and timber volume disturbed (Kautz et al., 2017; Seidl
172 et al., 2009). The model *inter alia* accounts for insect abundance, phenology and development,
173 as well as emergence and dispersal. It computes the number of beetle generations and sister
174 broods developed per year as well as winter survival rates based on the prevailing climate and
175 weather conditions, and considers individual tree defense capacity and susceptibility (simulated
176 via the non-structural carbohydrates pool of individual trees). Thus the model accounts for inter-
177 annual variation in the interactions between trees and bark beetles. Interactions between wind
178 and bark beetle disturbances arise from a high infestation probability and low defense capacity
179 of freshly downed trees after wind disturbance, while newly formed gaps (e.g., by bark beetles)
180 increase the exposure of surrounding forests to storm events. Seidl and Rammer (2017) found
181 that iLand is well able to reproduce these interactions for Kalkalpen National Park.

182 In addition to the submodules of natural disturbance we used the agent-based forest
183 management module (ABE) in iLand (Rammer and Seidl, 2015) to simulate past forest
184 management. ABE enables the dynamic application of generalized stand treatment programs,
185 including planting, tending, thinning, and harvesting activities. The dynamically simulated
186 management agent observes constraints at the stand and landscape scales, such as maximum

187 clearing sizes and sustainable harvest levels. Besides silvicultural treatments, we used ABE to
188 emulate the past management practice of salvage logging after bark beetle outbreaks.

189 iLand simulates a closed carbon cycle, tracking C in both aboveground (stem, branch, foliage,
190 tree regeneration) and belowground live tree compartments (coarse and fine roots).
191 Decomposition rates of detrital pools are modified by temperature and humidity to allow for
192 the simulation of C dynamics under changing climatic conditions. Detrital pools include litter
193 (i.e., dead material from both leaf and fine root turnover) and soil organic matter (Kätterer and
194 Andrén, 2001) as well as snags and downed coarse woody debris.

195 iLand has been extensively evaluated against independent data from forest ecosystems of the
196 northern front range of the Alps using a pattern-oriented modeling approach (Grimm et al.,
197 2005). The patterns for which simulations were compared against independent observations
198 include tree productivity gradients and natural vegetation dynamics (Thom et al., 2017b), wind
199 and bark beetle disturbance levels and distribution (Seidl and Rammer 2017), as well as
200 management trajectories (Albrich et al., 2018). A comprehensive documentation of iLand can
201 be found online at <http://iland.boku.ac.at/>, where also the model executable and source code are
202 freely available under a GNU GPL open source license.

203

204 **2.3 Reconstructing forest disturbance and land use history**

205 The study area has a long history of intensive timber harvesting for charcoal production, mainly
206 driven by a local pre-industrial iron-producing syndicate. This syndicate was active until 1889,
207 when the land was purchased by the k.k. (“kaiserlich und königlich”) Ministry for Agriculture.
208 During the 20th century, the majority of the landscape was managed by the Austrian Federal
209 Forests, and only limited areas within the landscape were still under the ownership of industrial
210 private companies (Weichenberger, 1994, 1995; Weinfurter, 2005). Forest management in the

211 late 19th and early 20th century was strongly influenced by the emerging industrialization. The
212 substitution of wood by mineral coal for heating, but especially for industrial energy supply,
213 changed the focus of forest management from fuel wood to timber production. At the same
214 time, an increase in agricultural productivity (also triggered by an input of fossil resources and
215 artificial fertilizer) allowed for the abandonment of less productive agricultural plots, often
216 followed by afforestation or natural regrowth of forest vegetation. Consequently, growing
217 stocks increased in many parts of Europe throughout the 20th century as the result of increases
218 in both forest extent and density (Bebi et al., 2017). In our study system, the shifting focus from
219 fuel wood to timber production around 1900 was accompanied by the introduction of systematic
220 stand delineation for spatial management planning (Fig. S1) as well as decadal inventories and
221 forest plan revisions. These documents are preserved in the archives of the Austrian Federal
222 Forests, and were used here to reconstruct past forest vegetation as well as management and
223 disturbance history (see Section S1, Fig. S1 and S2 in the Supplementary Material for details).

224 The oldest historic vegetation data available for the landscape were from an inventory
225 conducted between the years 1898 and 1911 and comprised growing stock and age classes for
226 11 tree species at the level of stand compartments for the entire landscape; we subsequently
227 used the year 1905 (representing the area-weighted mean year of this initial inventory) as the
228 temporal starting point for our analyses (Fig. 2). A major challenge for managers was to extract
229 resources from remote and inaccessible parts of the topographically highly complex landscape.

230 The most important means of timber transportation in the early 20th century was drifting (i.e.,
231 flushing logs down creeks and streams after artificially damming them). However, this
232 transportation technique was not feasible for heavy hardwood timber such as beech (Grabner et
233 al., 2004). Consequently, managers harvested trees selectively, and mainly focused on
234 accessible areas (i.e., stands close to streams). This resulted in some parts of the landscape

235 holding young, recently cut forests, while others containing stands of >160 years of age (Fig.
236 S3).

237 In addition to deriving the state of the forest in 1905, we reconstructed management activities
238 (thinnings, final harvests, artificial regeneration) and natural disturbances (wind and bark beetle
239 outbreaks) until 2013. From 1905 to 1917 timber extraction was fairly low. Between 1917 and
240 1923, however, a major disturbance episode by wind and bark beetles hit the region. Resulting
241 from a lack of labor force (military draft, malnutrition) in the last year of World War I a major
242 windthrow in 1917 could not be cleared, and the resulting bark beetle outbreak affected large
243 parts of the landscape. Overall, wind and bark beetles disturbed approximately one million
244 cubic meters of timber in the region between 1917 and 1923 (based on archival sources; Soyka,
245 1936; Weichenberger, 1994). Consequently, a railroad was installed to access and salvage the
246 disturbed timber. After the containment of the bark beetle outbreak in 1923 forest management
247 resumed at low intensity and no major natural disturbances were recorded. Following World
248 War II, a network of forest roads was built in order to gradually replace timber transportation
249 by railroads. The introduction of motorized chain saws (Fig. 2) further contributed to an
250 intensification of harvests. By 1971, forest railroads were completely replaced by motorized
251 transportation on forest roads, resulting in a further increase in the timber extracted from the
252 landscape. Timber removals from management as well as natural disturbances by wind and bark
253 beetles between 1905 and 1997 were reconstructed from annual management reviews available
254 from archival sources. With the landscape becoming part of KANP forest management ceased
255 in 1997. A second major natural disturbance episode affected the landscape from 2007-2013,
256 when a large bark beetle outbreak followed three storm events in 2007 and 2008. This second
257 disturbance episode was reconstructed from disturbance records of KANP in combination with
258 remote sensing data (Seidl and Rammer, 2016; Thom et al., 2017b).

259

260 **2.4 Landscape initialization and drivers**

261 The vegetation data for the year 1905 were derived from historical records for 2079 stands with
262 a median stand size of 1.7 ha. On average over the landscape, the growing stock was 212.3 m³
263 ha⁻¹ in 1905. The most common species were Norway spruce (with a growing stock of on
264 average 116.3 m³ ha⁻¹), European beech (68.0 m³ ha⁻¹), and European larch (*Larix decidua*
265 [Mill.], 21.5 m³ ha⁻¹). With an average growing stock of 4.2 m³ ha⁻¹ silver fir was considerably
266 underrepresented on the landscape relative to its role in the potential natural vegetation
267 composition, resulting from historic clear-cut management and high browsing pressure from
268 deer (see also Kučeravá et al., 2012). Despite these detailed records on past vegetation not all
269 information for initializing iLand were available from archival sources, e.g., diameters at breast
270 height (dbh) and height of individual trees, as well as tree positions, regeneration and
271 belowground carbon-pools had to be reconstructed by other means. To that end we developed
272 a new method for initializing vegetation and carbon pools in iLand, combining spin-up
273 simulations with empirical reference data on vegetation state, henceforth referred to as “legacy
274 spin-up”.

275 Commonly, spin-ups run models for a certain amount of time or until specified stopping criteria
276 are reached (e.g., steady-state conditions). The actual model-based analysis is then started from
277 the thus spun-up vegetation condition (Thornton and Rosenbloom, 2005). This has the
278 advantage that the model-internal dynamics (e.g., the relationships between the different C and
279 N pools in an ecosystem) are consistent when the focal analysis starts. However, the thus
280 derived initial vegetation condition frequently diverges from the vegetation state observed at a
281 given point in time (e.g., due to not all processes being represented in the applied model), and
282 does not account for the legacies of past management and disturbance. The legacy spin-up
283 approach developed here aims to reconstruct an (incompletely) known reference state of the
284 vegetation (e.g., the species composition, age, and growing stock reconstructed from archival

285 sources for the current analysis) from simulations (Fig. S4). To this end, iLand simulates long-
286 term forest development for each stand under past management and disturbance regimes.
287 During the simulations, the emerging forest trajectory is periodically compared to the respective
288 reference values, and the assumed past management is adapted iteratively in order to decrease
289 the difference between simulated vegetation states and observed reference values. This
290 procedure is executed in parallel for all stands on the landscape over a long period of time (here:
291 1000 years). The simulated vegetation state best corresponding to the reference values is stored
292 individually for each stand (including individual tree properties, regeneration, and carbon
293 pools), and later used to initialize model-based scenario analyses. A detailed description of the
294 legacy spin-up approach is given in the Supplementary Material Section S2.

295 In simulating 20th century forest dynamics we accounted for the abandonment of cattle grazing
296 and litter raking in forests (Glatzel, 1991) as well as an increasing atmospheric deposition of
297 nitrogen (Dirnböck et al., 2014; Roth et al., 2015). Specifically, we dynamically modified the
298 annual plant available nitrogen in our simulations based on data of nitrogen deposition in
299 Austria between 1880 and 2010, with nitrogen input peaking in the mid 1980s, followed by a
300 decrease and a stabilization after 2000 (Dirnböck et al., 2017). Besides edaphic factors also an
301 increase in temperature has led to more favorable conditions of tree growth (Pretzsch et al.,
302 2014). Detailed observations of climate for our study region reach back to 1950. Climate data
303 were statistically downscaled to a resolution of 100 × 100 m by means of quantile mapping,
304 accounting for topographic differences in climate conditions (Thom et al., 2017b). The lack of
305 detailed climate information before 1950 required an extension of the climate time series for
306 the years 1905 to 1949. To that end, we extracted data from the nearest weather station covering
307 the period from 1905 to present (i.e., Admont, located approximately 20 km south of our study
308 area), and used its temperature and precipitation record to sample years with corresponding
309 conditions from the observational record for our study landscape.

310 After using the legacy spin-up to generate tree vegetation and carbon pools in 1905, simulations
311 were run from 1905 until 2099, considering four different climate scenarios for the period 2013
312 – 2099. Climate change was represented by three combinations of global circulation models
313 (GCM) and regional climate models (RCM) under A1B forcing, including CNRM-RM4.5
314 (Radu et al., 2008) driven by the GCM ARPEGE, and MPI-REMO (Jacob, 2001) as well as
315 ICTP-RegCM3 (Pal et al., 2007), both driven by the GCM ECHAM5. The A1B scenario family
316 assumes rapid economic growth with global population peaking mid-century and declining
317 thereafter, and a balanced mix of energy sources being used (IPCC 2000). With average
318 temperature increases of between +3.1°C and +3.3°C and changing annual precipitation sums
319 of -87.0 mm to +135.6 mm by the end of the 21st century, the scenarios studied here are
320 comparable to the changes expected under the representative concentration pathways RCP4.5
321 and RCP6.0 for our study region (Thom et al., 2017c). In addition to the three scenarios of
322 climate change a historic baseline climate scenario was simulated. The years 1950 – 2010 were
323 used to represent this climatic baseline, and were randomly resampled to derive a stationary
324 climate time series until 2099.

325

326 **2.5 Analyses**

327 First, we evaluated the ability of iLand to reproduce the empirical data gathered for the studied
328 landscape. Following a pattern-oriented modeling approach (Grimm et al., 2005) we evaluated
329 a suit of different processes such as tree growth and competition, natural disturbances and forest
330 management. Specifically, we compared model outputs for different aspects of landscape
331 development (e.g., species composition, harvested and disturbed growing stock) at various
332 points in time against empirically derived historical data.

333 To address our first objective, i.e. investigating the spatio-temporal interactions of natural
334 disturbances, we used the empirically derived stand-level records of the two historic disturbance
335 episodes (1917 – 1923 and 2007 – 2013). We discretized the information (disturbed/
336 undisturbed) and rasterized the stand polygon data to a grid of 10×10 m. Subsequently, we
337 used this grid to calculate an odds ratio for the probability that the two disturbance events
338 affected the same locations on the landscape (i.e., the odds that areas disturbed in the first
339 episode were disturbed again in the second episode). We calculated the 95% confidence interval
340 of the odds ratio using the `vcd` package in R (Meyer et al., 2016).

341 To gain further insights into the drivers of the second disturbance period we ran simulations
342 under a combination of different land use and disturbance histories. Specifically, we
343 investigated the effect of two factors on the growing stock disturbed during the second
344 disturbance episode by controlling for their effects individually and in combination, resulting
345 in four simulated scenarios. The two factors considered were (i) the first episode of natural
346 disturbance (1917-1923), and (ii) forest management between 1923 (the end of the first
347 disturbance episode) and 1997 (the foundation of Kalkalpen National Park) (Fig. 2). Differences
348 among scenarios were compared by means of permutation-based independence tests using the
349 `coin` package (Hothorn et al., 2017).

350 To address our second objective, i.e., evaluating the impact of past land use and natural
351 disturbance as well as future climate on the 21st century carbon sink strength, we extended our
352 factorial simulation design to also account for the second disturbance episode and different
353 future climate scenarios. Hence, a third factor considered in the simulated landscape history
354 was the second natural disturbance episode (2007-2013) (Fig. 2). The factorial combination of
355 elements representing the actual history of our study landscape was chosen as a reference for
356 assessing the effects of past disturbance and land use on future C uptake. After 2013 four

357 different climate scenarios were simulated for all alternative disturbance histories, to assess the
358 impacts of climate change on the future NEE of the landscape.

359 All simulations were started from the landscape conditions in 1905, determined by means of
360 the legacy spin-up procedure described above. From 1905 to 1923 management and natural
361 disturbances were implemented in the simulation as recorded in the stand-level archival
362 sources. After 1923, natural disturbances were simulated dynamically using the respective
363 iLand disturbance modules. For the second disturbance episode (2007 – 2013) the observed
364 peak wind speeds for the storms Kyrill (2007), Emma (2008) and Paula (2008) were used in the
365 simulation (see Seidl and Rammer 2017 for details). Beyond 2013, natural disturbances were
366 dynamically simulated with iLand, however, we excluded high intensity wind disturbance
367 events to control for confounding effects with past disturbance events. Specifically, we
368 randomly sampled annual peak wind speeds from the distribution of years before 2006, and
369 simulated the wind and bark beetle dynamics emerging on the landscape (see also Thom et al.,
370 2017a).

371 Management interventions from 1924 to 1997 were simulated using ABE. The individual
372 silvicultural decisions were thus implemented dynamically by the management agent in the
373 model, based on generic stand treatment programs of past management in Austria's federal
374 forests and the emerging state of the forest. The advantage of this approach was that
375 management was realistically adapted to different forest states in the simulations, e.g., with
376 harvesting patterns differing in the runs in which the disturbance episode 1917 – 1923 was
377 omitted. Moreover, in line with the technical revolutions of the 20th century (Fig. 2) the
378 simulated management agent was set to account for an intensification of forest management
379 over time (e.g., a higher number of thinnings and shorter rotation periods). In summary, our
380 simulation design consisted of 32 combinations of different land use and disturbance histories
381 and climate futures (first disturbance episode (yes/no) × management (yes/no) × second

382 disturbance episode (yes/no) \times 4 climate scenarios). In order to account for the stochasticity of
383 iLand (e.g., with regard to bark beetle dispersal distance and direction, uprooting and breakage
384 probability during storm events etc.) we replicated each scenario combination 20 times (i.e., in
385 total 640 simulation runs) for the years 1905 – 2099 (195 years).

386 We evaluated the ability of iLand to reproduce past natural disturbance and land use as well as
387 the resultant forest vegetation dynamics on the landscape by comparing simulations of the
388 baseline scenario (i.e., including historic climate, as well as reconstructed natural disturbance
389 and land use) with independent empirical data for different time periods: the simulated amount
390 of timber extracted was compared to historical records for three time periods signifying major
391 technical system changes during the 20th century (Fig. 2). Simulated impacts of the second
392 disturbance episode (2007 – 2013) on growing stock were compared against empirical records
393 from KANP. Model outputs for species shares and total growing stock were compared against
394 historical records for the year 1905, testing the ability of the legacy spin-up to recreate the initial
395 vegetation state. Furthermore, simulated species shares and growing stocks were related to
396 observations for 1999, i.e., testing the capacity of iLand to faithfully reproduce forest conditions
397 after 95 years of vegetation dynamics. The results of all these tests can be found in the
398 Supplement Sections S2 and S3.

399 We used simulation outputs to investigate the changes in NEE over time and across different
400 scenarios. NEE denotes the net C flux from the ecosystem to the atmosphere, with negative
401 values indicating ecosystem C gain (Chapin et al., 2006). To determine the impact of past
402 disturbance and land use as well as future climate on the 21st century carbon balance of the
403 landscape, we first computed the cumulative NEE over the period 2014 – 2099 for each
404 simulation (i.e., after land use ceased and the two disturbance episodes were over in order to
405 enable the analysis of their future effects on NEE). Next, the effects of past disturbance and
406 land use as well as future climate were determined from mean differences between the different

407 factor combinations in the simulation experiment with regard to their cumulative NEE in 2099.
408 P-values were computed by means of independence tests (Hothorn et al., 2017). All analyses
409 were performed using the R language and environment for statistical computing (R
410 Development Core Team 2017).

411

412 **3. Results**

413 **3.1 Reconstructing historic landscape dynamics**

414 Using iLand, we were able to successfully reproduce historic vegetation dynamics on the
415 landscape. The species composition of the legacy spin-up diverged by 2.3% (weighted by the
416 observed growing stock), while the simulated growing stock was on average 4.6% lower than
417 the reference state in 1905 (see Section S2 including Fig. S5, Fig. S6). Furthermore, the iLand
418 management module ABE reproduced the intensification of forest management over the 20th
419 century close to the observed values (average divergence: $+0.2 \text{ m}^3 \text{ ha}^{-1} \text{ yr}^{-1}$) (Fig. S7). The first
420 evaluation period (1924 – 1952) resulted in a slightly larger overestimation of simulated
421 harvests (on average $+0.3 \text{ m}^3 \text{ ha}^{-1} \text{ yr}^{-1}$). Further, the simulated wind and bark beetle disturbances
422 between 2007 and 2013 corresponded well to the expected values derived from KANP
423 inventories (divergence: $-0.1 \text{ m}^3 \text{ ha}^{-1}$) (Fig. S8). Our dynamic simulation approach adequately
424 reproduced the tree species composition (average deviation of 2.5% in species shares weighted
425 by observed growing stock) and growing stock ($+9.2 \text{ m}^3 \text{ ha}^{-1}$) at the landscape scale after 95
426 years of simulation (Fig. S9). Despite an intensification of harvests until 1997 and the
427 occurrence of a major disturbance event in 1917 – 1923, the average growing stock on the
428 landscape doubled between 1905 and 2013 (Fig. S10). At the same time total ecosystem carbon
429 increased by 40.9% (Fig. S11). European beech dominance increased over the 20th century, in

430 particular at lower elevations (Fig. S10, Fig. 1e and 1f). Further details on historic landscape
431 development can be found in the Supplement in Sections S2 and S3 (Fig. S4-S11).

432

433 **3.2 Long-term drivers of natural disturbances**

434 We used the empirically derived spatial footprint of two episodes of natural disturbance 90
435 years apart to investigate the long-term temporal interactions between disturbances. Both
436 disturbance episodes were found to have a similar impact on growing stock (117,441 m³ and
437 93,084 m³ of growing stock disturbed, respectively), whereas the first episode affected an area
438 more than twice the size of the second episode (2334 ha and 1116 ha, respectively). Only 9.2%
439 of the area disturbed during the first episode was also affected by the second episode (Fig. 3).
440 Whereas the first disturbance episode mainly affected the central and southern reaches of the
441 study area, the effects of the second disturbance episode were most pronounced in the northern
442 parts of the landscape. The odds ratio of 0.49 ($p < 0.001$) revealed a lower probability that the
443 same location of the first disturbance episode is affected by the second disturbance episode on
444 the landscape compared to the odds that a previously undisturbed area is disturbed by the second
445 disturbance episode. Based on our simulations we found only a moderate positive effect of the
446 first disturbance episode on the volume disturbed during the second episode (+8,181 m³,
447 $p = 0.401$). In contrast, land use had a considerable impact on the second disturbance episode.
448 On average, land use increased the volume disturbed by +28.927 m³ ($p < 0.001$).

449

450 **3.3 The effect of past disturbance and land use as well as future climate on** 451 **21st century carbon sequestration**

452 Our simulations revealed a considerable impact of past land use on the current state of total
453 ecosystem carbon (Table 1). On average over all scenarios, the cessation of land use resulted in
454 an increase in carbon stocks of $+39.7 \text{ tC ha}^{-1}$ (+9.2%) in 2013. The two episodes of natural
455 disturbance had a limited effect on current carbon stocks. The omission of both natural
456 disturbance episodes increased carbon stocks in 2013 by only $+4.2 \text{ tC ha}^{-1}$ (+0.9%). Conversely,
457 past land use initiated a strong and continuous positive legacy effect on the future cumulative
458 carbon uptake of the landscape beyond 2013 (Table 1, Fig. 4), resulting from a persistent
459 recovery of growing stocks (Table 2). Notably, past land use caused a cumulative decrease in
460 future NEE of -41.8 tC ha^{-1} ($p < 0.001$) until 2099 on average over all scenarios. The second
461 disturbance episode resulted in an initial release of carbon (positive NEE) lasting for several
462 years after the event, followed by a reversal of the trend towards a negative NEE effect (Fig.
463 4). Its overall impact on cumulative NEE at the end of the simulation period was -3.1 tC ha^{-1}
464 ($p = 0.191$), i.e. over the 21st century the recent disturbance period had an overall positive effect
465 on forest C sequestration. The first disturbance episode (1917-1923) had almost no effect on
466 the forest carbon dynamics in the 21st century (NEE effect of -0.6 tC ha^{-1} , $p = 0.792$).

467 Climate change weakened the carbon sink strength on the landscape, mainly as a result of a
468 climate-mediated alteration of successional trajectories (Table 2). Driven by a strong reduction
469 of Norway spruce (on average $-46.4 \text{ m}^3 \text{ ha}^{-1}$), the growing stock on the landscape was on
470 average $9.3 \text{ m}^3 \text{ ha}^{-1}$ lower in comparison to simulations with historic climate. Also, climate
471 change effects on NEE were more variable and increased in uncertainty over time as a result of
472 differences in climate scenarios (mean 464.1 tC ha^{-1} ; SD 21.0 tC ha^{-1}) compared to land-use
473 (mean 488.3 tC ha^{-1} ; SD 0.9 tC ha^{-1}) and disturbance legacy effects (mean 487.2 tC ha^{-1} ; SD
474 1.0 tC ha^{-1}) (Table 1, Fig. 4). On average, climate change increased the cumulative NEE until
475 2099 by $+22.9 \text{ tC ha}^{-1}$ ($p < 0.001$), and thus reduced the carbon uptake of the landscape relative
476 to a continuation of historic climate (Fig. 4).

477

478 **4. Discussion**

479 **4.1 Human and natural disturbance interactions**

480 Based on previous studies assessing the spatial and temporal autocorrelation of disturbances in
481 Europe (Marini et al., 2012; Schurman et al., 2018; Stadelmann et al., 2013; Thom et al., 2013)
482 we hypothesized that a disturbance episode in the early 20th century influenced disturbances in
483 the early 21st century. However, our analysis revealed a low probability for the same area to be
484 affected by two consecutive disturbance episodes of the same disturbance agents (Fig. 3).
485 Moreover, our simulations only indicate a weak correlation between the two consecutive
486 disturbance episodes on the landscape. Hence, our data do not support the hypothesis of
487 amplified disturbance interactions and long-term cyclic disturbance in Central European
488 forests. Our initial assumption was based on the expectation of uniform recovery after the first
489 disturbance episode, with large parts of the landscape reaching high susceptibility to wind and
490 bark beetles simultaneously. However, disturbances can also have negative, dampening effects
491 on future disturbance occurrence, e.g., when they lead to increased heterogeneity (Seidl et al.,
492 2016) and trigger autonomous adaptation of forests to novel environmental conditions (Thom
493 et al., 2017c). The low overlap between the two disturbance episodes reported here could thus
494 be an indication for such a dampening feedback between disturbances in parts of the landscape,
495 yet further tests are needed to substantiate this hypothesis for Central European forest
496 ecosystems. An alternative explanation for the diverging spatial patterns of the two disturbance
497 episodes might be a different wind direction in the storm events initiating the two respective
498 episodes, affecting different parts of the highly complex mountain forest landscapes. Also the
499 legacy effects from past land use were different for each episode. The more open structure
500 within stands resulting from heavy exploitation before 1900 may, for instance, have increased

501 wind susceptibility in the central and southern reaches of the landscape. These diverging
502 hypotheses of dampening effects between sequential disturbance episodes after several decades
503 of forest recovery should be tested in a factorial simulation experiment in the future (e.g.,
504 assessing the effects of disturbance-induced forest heterogeneity on a subsequent disturbance
505 episode or testing the effects of different wind directions of sequential disturbance episodes).

506 In contrast to our finding regarding interactions between natural disturbances, our simulations
507 supported our expectation of an amplifying effect of past land use on recent disturbance activity.
508 This finding is congruent with other analyses suggesting past forest management as a driver of
509 current natural disturbance regimes (Hanewinkel et al., 2014; Schelhaas, 2008; Seidl et al.,
510 2011). Past forest management in Central Europe has, for instance, strongly promoted Norway
511 spruce, which is one of the most vulnerable species to natural disturbances in the region
512 (Hanewinkel et al., 2008; Pasztor et al., 2014). Pure stands of Norway spruce are particularly
513 conducive to large-scale eruptions of bark beetles, and even-aged management creates edges
514 that are highly susceptible to strong winds (Hanewinkel et al., 2014; Thom et al., 2013). Our
515 analysis thus suggests that as disturbances increase under climate change (Seidl et al., 2017;
516 Thom et al., 2017a), forests that have been homogenized by past land use are at particular risk.

517

518 **4.2 The role of legacies on future C uptake**

519 Past studies investigating drivers of the forest carbon balance have largely focused either on
520 historic factors (Keenan et al., 2014; Naudts et al., 2016) or future changes in the environment
521 (Manusch et al., 2014; Reichstein et al., 2013). Only few studies to date have explicitly
522 quantified the effect of legacies from natural disturbance and land use when assessing climate
523 change impacts on the future carbon uptake of forest ecosystems. However, disregarding legacy
524 effects could lead to a misattribution of future forest C changes. Here we harnessed an extensive

525 long-term documentation of vegetation history to study impacts of past natural disturbance and
526 land use as well as future climate on the future NEE of a forest landscape. We found long-
527 lasting legacy effects of both past natural disturbance and land use on the forest carbon cycle
528 (see also Gough et al., 2007; Kashian et al., 2013; Landry et al., 2016; Nunery and Keeton,
529 2010), supporting our hypothesis regarding the importance of legacies for future C dynamics.
530 While the legacy effect of past land use was strong, the impact of natural disturbances on the
531 future NEE was an order of magnitude lower (Fig. 4). Here it is important to note that our results
532 are strongly contingent on the intense and century-long land use history in Central Europe. A
533 dynamic landscape simulation study for western North America, for instance, emphasized the
534 dominant role of natural disturbances to determine future NEE (Loudermilk et al., 2013). In our
535 study system, however, land use legacies may have a stronger effect on future NEE than past
536 natural disturbances and future changes in climatic conditions (e.g., in our study area forest
537 management altered forests more strongly than natural disturbances in most years) (Fig. 4).
538 Disregarding legacy effects may thus cause a substantial bias when studying the future carbon
539 dynamics of forest ecosystems. It has to be noted, however, that our study only considered three
540 relatively moderate climate change scenarios. Hence we might underestimate the effect of
541 climate change on NEE, if future climate change will follow a more severe trajectory (see e.g.,
542 Kruhlov et al, 2018). Furthermore, it is likely that over longer future time frames as the one
543 studied here the effects of climate change will become more important relative to past legacy
544 effects (Temperli et al., 2013).

545 While we here focused on the strength of legacy effects, our results also provide insights into
546 their duration. Land-use related differences in C stocks persisted throughout the simulation
547 period, with trajectories converging only towards the end of the 21st century. Hence, our data
548 indicate that land use legacies affect the forest C cycle for at least one century in our study
549 system. Despite the considerably lower impacts of natural disturbances, the legacy effect of the

550 second disturbance episode also lasted for several decades (Fig. 4). Future efforts should aim
551 at determining the duration of past legacies more precisely, considering a variety of different
552 forest conditions (e.g., Temperli et al., 2013). Moreover, while we here focus on the effects of
553 wind and bark beetle disturbances – currently the two most important natural disturbance agents
554 in Central Europe (Thom et al., 2013) – as well as their interactions, future climate change may
555 increase the importance of other disturbance agents not investigated here (see e.g., Wingfield
556 et al., 2017).

557 The specific disturbance history of our study area, characterized by an intensive disturbance
558 and land use history and major socio-ecological transitions throughout the 20th century, is key
559 for interpreting our findings. In particular, the cessation of forest management in 1997 had a
560 very strong impact on the future carbon balance of the landscape (an on average 52.8 and 13.4
561 times higher effect than the first and second episodes of natural disturbances, respectively – see
562 Fig. 4). In addition to disturbance legacy effects, also climate change significantly affected the
563 future NEE. In contrast to the general notion that temperate forests will serve as a strong carbon
564 sink under climate change (Bonan, 2008), our dynamic simulations suggest that climate change
565 will decrease the ability of the landscape to sequester carbon in the future, mainly by forcing a
566 transition to forest types with a lower carbon storage potential (see also Kruhlov et al., 2018;
567 Thom et al., 2017a). However, considerable uncertainties of climate change impacts on the
568 carbon balance of forest ecosystems remain (e.g., Manusch et al., 2014). These uncertainties
569 may arise from a wide range of potential future climate trajectories, but also from a limited
570 understanding of processes such as the CO₂ fertilization effect on forest C uptake (Kroner and
571 Way, 2016; Reyer et al., 2014). In addition to the direct impacts of climate change (e.g., via
572 temperature and precipitation changes) on forest ecosystems, climate change will also alter
573 future natural disturbance regimes (Seidl et al., 2017). The potential for such large pulses of C

574 release from forests is rendering the role of forests in climate mitigation strategies highly
575 uncertain (Kurz et al., 2008; Seidl et al., 2014a).

576

577 **5. Conclusions**

578 Past natural disturbance regimes and land use have a long-lasting influence on forest dynamics.

579 In order to project the future of forest ecosystems we thus need to better understand their past.

580 We here showed how a combination of historical sources and simulation modeling – applied

581 by an interdisciplinary team of scientists – can be used to improve our understanding of the

582 long-term trajectories of forest ecosystems (Bürgi et al., 2017; Collins et al., 2017; Deng and

583 Li, 2016). Two conclusions can be drawn from the strong historical determination of future

584 forest dynamics: First, as temperate forests have been managed intensively in many parts of the

585 world (Deng and Li, 2016; Foster et al., 1998; Naudts et al., 2016), their contribution to climate

586 change mitigation over the coming decades is likely determined already to a large degree by

587 their past (see also Schwaab et al., 2015). This means that for the time frame within which a

588 transformation of human society needs to be achieved in order to retain the earth system within

589 its planetary boundaries (Steffen et al., 2011), the potential for influencing the role of forests

590 might be lower than frequently assumed. Efforts to change forest management now to mitigate

591 climate change through *in situ* C storage have high potential (Canadell and Raupach, 2008), but

592 will likely unfold their effects too late to make a major contribution to climate mitigation in the

593 coming decades. Second, any intentional (by forest management) or unintentional (by natural

594 disturbances) changes in forest structure and composition may have profound consequences for

595 the future development of forest ecosystems. This underlines that a long-term perspective

596 integrating past and future ecosystem dynamics is important when studying forests, and that

597 decadal to centennial foresight is needed in ecosystem management.

598

599 **Author contribution**

600 RS, DT and WR designed the study, RG collected historical data from archives, DT and WR
601 performed simulations, DT analyzed the outputs, all authors contributed to writing the
602 manuscript.

603

604 **Competing interests**

605 The authors declare that they have no conflict of interest.

606

607 **Acknowledgements**

608 This study was supported by the Austrian Climate and Energy Fund ACRP (grant
609 KR14AC7K11960). W. Rammer and R. Seidl acknowledge further support from the Austrian
610 Science Fund FWF through START grant Y895-B25. We thank the Austrian Federal Forests
611 for the permission to access their archives for the collection of the historic data used in this
612 study. We also acknowledge the comments of three anonymous reviewers which helped greatly
613 to improve our study. The simulations performed in this study were conducted at the Vienna
614 Scientific Cluster.

615

616 **References**

617 Albrich, K., Rammer, W., Thom, D., Seidl, R.: Trade-offs between temporal stability and level
618 of forest ecosystem services provisioning under climate change, in press., 2018.

619 Bebi, P., Seidl, R., Motta, R., Fuhr, M., Firm, D., Krumm, F., Conedera, M., Ginzler, C.,
620 Wohlgemuth, T. and Kulakowski, D.: Changes of forest cover and disturbance regimes in
621 the mountain forests of the Alps, *For. Ecol. Manage.*, 388, 43–56,
622 doi:10.1016/j.foreco.2016.10.028, 2017.

623 Bonan, G. B.: Forests and climate change: forcings, feedbacks, and the climate benefits of
624 forests., *Science*, 320(5882), 1444–1449, doi:10.1126/science.1155121, 2008.

625 Bürgi, M., Östlund, L. and Mladenoff, D. J.: Legacy effects of human land use: Ecosystems as
626 time-lagged systems, *Ecosystems*, 20(1), 94–103, doi:10.1007/s10021-016-0051-6, 2017.

627 Canadell, J. G. and Raupach, M. R.: Managing forests for climate change mitigation, *Science*,
628 320(5882), 1456–1457, doi:DOI 10.1126/science.1155458, 2008.

629 Chapin, F. S., Woodwell, G. M., Randerson, J. T., Rastetter, E. B., Lovett, G. M., Baldocchi,
630 D. D., Clark, D. A., Harmon, M. E., Schimel, D. S., Valentini, R., Wirth, C., Aber, J. D.,
631 Cole, J. J., Goulden, M. L., Harden, J. W., Heimann, M., Howarth, R. W., Matson, P. A.,
632 McGuire, A. D., Melillo, J. M., Mooney, H. A., Neff, J. C., Houghton, R. A., Pace, M.
633 L., Ryan, M. G., Running, S. W., Sala, O. E., Schlesinger, W. H. and Schulze, E. D.:
634 Reconciling carbon-cycle concepts, terminology, and methods, *Ecosystems*, 9(7), 1041–
635 1050, doi:10.1007/s10021-005-0105-7, 2006.

636 Collins, B. M., Fry, D. L., Lydersen, J. M., Everett, R. and Stephens, S. L.: Impacts of
637 different land management histories on forest change, *Ecol. Appl.*, 0(0), 1–12,
638 doi:10.1002/eap.1622, 2017.

639 Deng, X. and Li, Z.: A review on historical trajectories and spatially explicit scenarios of
640 land-use and land-cover changes in China, *J. Land Use Sci.*, 11(6), 709–724,
641 doi:10.1080/1747423X.2016.1241312, 2016.

642 Dirnböck, T., Grandin, U., Bernhardt-Römermann, M., Beudert, B., Canullo, R., Forsius, M.,
643 Grabner, M.-T., Holmberg, M., Kleemola, S., Lundin, L., Mirtl, M., Neumann, M.,
644 Pompei, E., Salemaa, M., Starlinger, F., Staszewski, T. and Uziębło, A. K.: Forest floor
645 vegetation response to nitrogen deposition in Europe, *Glob. Chang. Biol.*, 20(2), 429–
646 440, doi:10.1111/gcb.12440, 2014.

647 Dirnböck, T., Djukic, I., Kitzler, B., Kobler, J., Mol-Dijkstra, J. P., Posch, M., Reinds, G. J.,
648 Schlutow, A., Starlinger, F. and Wamelink, W. G. W.: Climate and air pollution impacts
649 on habitat suitability of Austrian forest ecosystems, edited by R. Zang, *PLoS One*, 12(9),
650 e0184194, doi:10.1371/journal.pone.0184194, 2017.

651 Drake, J. E., Gallet-Budynek, A., Hofmockel, K. S., Bernhardt, E. S., Billings, S. A., Jackson,
652 R. B., Johnsen, K. S., Lichter, J., Mccarthy, H. R., McCormack, M. L., Moore, D. J. P.,
653 Oren, R., Palmroth, S., Phillips, R. P., Phippen, J. S., Pritchard, S. G., Treseder, K. K.,
654 Schlesinger, W. H., Delucia, E. H. and Finzi, A. C.: Increases in the flux of carbon
655 belowground stimulate nitrogen uptake and sustain the long-term enhancement of forest
656 productivity under elevated CO₂, *Ecol. Lett.*, 14(4), 349–357, doi:10.1111/j.1461-
657 0248.2011.01593.x, 2011.

658 Erb, K.-H.: Land use related changes in aboveground carbon stocks of Austria's terrestrial
659 ecosystems, *Ecosystems*, 7(5), 563–572, doi:10.1007/s10021-004-0234-4, 2004.

660 Erb, K. H., Kastner, T., Plutzer, C., Bais, A. L. S., Carvalhais, N., Fetzel, T., Gingrich, S.,
661 Haberl, H., Lauk, C., Niedertscheider, M., Pongratz, J., Thurner, M. and Luysaert, S.:
662 Unexpectedly large impact of forest management and grazing on global vegetation
663 biomass, *Nature*, 553(7686), 73–76, doi:10.1038/nature25138, 2018.

664 Foster, D. R., Motzkin, G. and Slater, B.: Land-Use History as Long-Term Broad-Scale
665 Disturbance: Regional Forest Dynamics in Central New England, *Ecosystems*, 1(1), 96–

666 119, doi:10.1007/s100219900008, 1998.

667 Glatzel, G.: the Impact of Historic Land-Use and Modern Forestry on Nutrient Relations of
668 Central-European Forest Ecosystems, *Fertil. Res.*, 27(1), 1–8, doi:10.1007/BF01048603,
669 1991.

670 Goetz, S. J., Bond-Lamberty, B., Law, B. E., Hicke, J. A., Huang, C., Houghton, R. A.,
671 McNulty, S., O’Halloran, T., Harmon, M., Meddens, A. J. H., Pfeifer, E. M., Mildrexler,
672 D. and Kasischke, E. S.: Observations and assessment of forest carbon dynamics
673 following disturbance in North America, *J. Geophys. Res. Biogeosciences*, 117(2), 1–17,
674 doi:10.1029/2011JG001733, 2012.

675 Gough, C. M., Vogel, C. S., Harrold, K. H., George, K. and Curtis, P. S.: The legacy of
676 harvest and fire on ecosystem carbon storage in a north temperate forest, *Glob. Chang.*
677 *Biol.*, 13(9), 1935–1949, doi:10.1111/j.1365-2486.2007.01406.x, 2007.

678 Grabner, M., Wimmer, R. and Weichenberger, J.: Reconstructing the History of Log-Drifting
679 in the Reichraminger Hintergebirge, Austria. 21, no. 3: 131-137., *Dendrochronologia*,
680 21(3), 131–137, 2004.

681 Grimm, V.E., Revilla E., Berger U., Jeltsch F., Mooij W.M., Railsback S.F., Thulke H.-H.,
682 Weiner J., Wiegand T., DeAngelis D.L.: Pattern-Oriented Modeling of Agent-Based
683 Complex Systems: Lessons from Ecology, *Science.*, 310(5750), 987–991,
684 doi:10.1126/science.1116681, 2005.

685 Hanewinkel, M., Breidenbach, J., Neeff, T. and Kublin, E.: Seventy-seven years of natural
686 disturbances in a mountain forest area — the influence of storm, snow, and insect damage
687 analysed with a long-term time series, *Can. J. For. Res.*, 38(8), 2249–2261,
688 doi:10.1139/X08-070, 2008.

689 Hanewinkel, M., Cullmann, D. A., Schelhaas, M.-J., Nabuurs, G.-J. and Zimmermann, N. E.:
690 Climate change may cause severe loss in the economic value of European forest land,
691 Nat. Clim. Chang., 3(3), 203–207, doi:10.1038/nclimate1687, 2012.

692 Hanewinkel, M., Kuhn, T., Bugmann, H., Lanz, A. and Brang, P.: Vulnerability of uneven-
693 aged forests to storm damage, Forestry, 87(4), 525–534, doi:10.1093/forestry/cpu008,
694 2014.

695 Harmon, M. E., Ferrel, W. K. and Franklin, J. F.: Effects on carbon storage of conversion of
696 old-growth forests to young forests, Science, 247, 699–702, 1990.

697 Hothorn, T., Hornik, K., van de Wiel, M. A., Winell, H., Zeileis, A.: Package 'coin'.
698 <https://cran.r-project.org/web/packages/coin/coin.pdf>, 2017.

699 IPCC: Special report on emission scenarios. Contribution of Working Group III of the
700 Intergovernmental Panel on Climate Change. In N. Nakicenovic, R. Swart (Eds.).
701 Emissions scenarios (pp. 1–570). Cambridge, UK: Cambridge University Press, 2000.

702 IPCC: Climate Change 2007: The Physical Science Basis. Contribution of Working Group I
703 to the Fourth Assessment Report of the Intergovernmental Panel on Climate Change. In
704 S. Solomon, D. Qin, M. Manning, Z. Chen, M. Marquis, K. B. Averyt, M. Tignor, H.
705 Miller (Eds.). Climate Change 2007: The Physical Science Basis (pp. 1–996).
706 Cambridge, UK: Cambridge University Press, 2007.

707 IPCC: Climate Change 2014: Mitigation of Climate Change. Contribution of Working Group
708 III to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change. In
709 O. R. Edenhofer, R. Pichs-Madruga, Y. Sokona, E. Farahani, S. Kadner, K. Seyboth, A.
710 Adler, I. Baum, S. Brunner, P. Eickemeier, B. Kriemann, J. Savolainen, S. Schlömer, C.
711 von Stechow, T. Zwickel, J. C. Minx (Eds.). Climate Change 2014: Mitigation of

712 Climate Change (pp. 1–1435). Cambridge, UK and New York, NY, USA: Cambridge
713 University Press, 2014.

714 Jacob, D.: A note to the simulation of the annual and inter-annual variability of the water
715 budget over the Baltic Sea drainage basin, *Meteorol. Atmos. Phys.*, 77(1–4), 61–73,
716 doi:10.1007/s007030170017, 2001.

717 Kashian, D. M., Romme, W. H., Tinker, D. B., Turner, M. G. and Ryan, M. G.: Postfire
718 changes in forest carbon storage over a 300-year chronosequence of *Pinus contorta* -
719 dominated forests, *Ecol. Monogr.*, 83(1), 49–66, doi:10.1890/11-1454.1, 2013.

720 Kätterer, T. and André, O.: The ICBM family of analytically solved models of soil carbon,
721 nitrogen and microbial biomass dynamics - Descriptions and application examples, *Ecol.*
722 *Modell.*, 136(2–3), 191–207, doi:10.1016/S0304-3800(00)00420-8, 2001.

723 Kautz, M., Meddens, A. J. H., Hall, R. J. and Arne, A.: Biotic disturbances in Northern
724 Hemisphere forests - a synthesis of recent data, uncertainties and implications for forest
725 monitoring and modelling, *Glob. Ecol. Biogeogr.*, 26(5), 533–552,
726 doi:10.1111/geb.12558, 2017.

727 Keenan, T. F., Gray, J., Friedl, M. A., Toomey, M., Bohrer, G., Hollinger, D. Y., Munger, J.
728 W., O’Keefe, J., Schmid, H. P., Wing, I. S., Yang, B. and Richardson, A. D.: Net carbon
729 uptake has increased through warming-induced changes in temperate forest phenology,
730 *Nat. Clim. Chang.*, 4(7), 598–604, doi:10.1038/nclimate2253, 2014.

731 Keenan, T. F., Prentice, I. C., Canadell, J. G., Williams, C. A., Wang, H., Raupach, M. and
732 Collatz, G. J.: Recent pause in the growth rate of atmospheric CO₂ due to enhanced
733 terrestrial carbon uptake, *Nat. Commun.*, 7, 13428, doi:10.1038/ncomms13428, 2016.

734 Kobler, J.: Risikokarten als Planungsgrundlage für Flächenbewirtschaftung und

735 Tourismuslenkung im Nationalpark Kalkalpen Oberösterreich. Vienna, Austria: Faculty
736 of Earth Sciences, Geography and Astronomy, University of Vienna, 2004.

737 Kroner, Y. and Way, D. A.: Carbon fluxes acclimate more strongly to elevated growth
738 temperatures than to elevated CO₂ concentrations in a northern conifer, *Glob. Chang.*
739 *Biol.*, 22(8), 2913–2928, doi:10.1111/gcb.13215, 2016.

740 Kruhlov, I., Thom, D., Chaskovskyy, O., Keeton, W. S. and Scheller, R. M.: Future forest
741 landscapes of the Carpathians: vegetation and carbon dynamics under climate change,
742 *Reg. Environ. Chang.*, 18(5), 1555–1567, doi:10.1007/s10113-018-1296-8, 2018.

743 Kučeravá, B., Dobrovolný, L. and Remeš, J.: Responses of *Abies alba* seedlings to different
744 site conditions in *Picea abies* plantations, *Dendrobiology*, 69, 49–58,
745 doi:10.12657/denbio.069.006, 2012.

746 Kurz, W. A., Stinson, G., Rampley, G. J., Dymond, C. C. and Neilson, E. T.: Risk of natural
747 disturbances makes future contribution of Canada ' s forests to the global carbon cycle
748 highly uncertain, *PNAS*, 105(5), 1551–1555, 2008.

749 Landry, J.-S., Parrott, L., Price, D. T., Ramankutty, N. and Matthews, H. D.: Modelling long-
750 term impacts of mountain pine beetle outbreaks on merchantable biomass, ecosystem
751 carbon, albedo, and radiative forcing, *Biogeosciences*, 13(18), 5277–5295,
752 doi:10.5194/bg-13-5277-2016, 2016.

753 Landsberg, J. J. and Waring, R. H.: A generalised model of forest productivity using
754 simplified concepts of radiation-use efficiency, carbon balance and partitioning, *For.*
755 *Ecol. Manage.*, 95(3), 209–228, doi:10.1016/S0378-1127(97)00026-1, 1997.

756 Loudermilk, E. L., Scheller, R. M., Weisberg, P. J., Yang, J., Dilts, T. E., Karam, S. L. and
757 Skinner, C.: Carbon dynamics in the future forest: The importance of long-term

758 successional legacy and climate-fire interactions, *Glob. Chang. Biol.*, 19(11), 3502–3515,
759 doi:10.1111/gcb.12310, 2013.

760 Manusch, C., Bugmann, H. and Wolf, A.: The impact of climate change and its uncertainty on
761 carbon storage in Switzerland, *Reg. Environ. Chang.*, 14(4), 1437–1450,
762 doi:10.1007/s10113-014-0586-z, 2014.

763 Marini, L., Ayres, M. P., Battisti, A. and Faccoli, M.: Climate affects severity and altitudinal
764 distribution of outbreaks in an eruptive bark beetle, *Clim. Change*, 115(2), 327–341,
765 doi:10.1007/s10584-012-0463-z, 2012.

766 Meyer, D., Zeileis, A., Hornik, K., Gerber, F., Friendly, M.: Package 'vcd'. [https://cran.r-](https://cran.r-project.org/web/packages/vcd/vcd.pdf)
767 [project.org/web/packages/vcd/vcd.pdf](https://cran.r-project.org/web/packages/vcd/vcd.pdf), 2016.

768 Naudts, K., Chen, Y., McGrath, M. J., Ryder, J., Valade, A., Otto, J. and Luysaert, S.:
769 Europes forest management did not mitigate climate warming, *Science*, 351(6273), 597–
770 600, doi:10.1126/science.aad7270, 2016.

771 Nunery, J. S. and Keeton, W. S.: Forest carbon storage in the northeastern United States: Net
772 effects of harvesting frequency, post-harvest retention, and wood products, *For. Ecol.*
773 *Manage.*, 259(8), 1363–1375, doi:10.1016/j.foreco.2009.12.029, 2010.

774 Overbeck, M. and Schmidt, M.: Modelling infestation risk of Norway spruce by *Ips*
775 *typographus* (L.) in the Lower Saxon Harz Mountains (Germany), *For. Ecol. Manage.*,
776 266, 115–125, doi:10.1016/j.foreco.2011.11.011, 2012.

777 Pal, J. S., Giorgi, F., Bi, X., Elguindi, N., Solmon, F., Rauscher, S. A., Gao, X., Francisco, R.,
778 Zakey, A., Winter, J., Ashfaq, M., Syed, F. S., Sloan, L. C., Bell, J. L., Diffenbaugh, N.
779 S., Karmacharya, J., Konaré, A., Martinez, D., da Rocha, R. P. and Steiner, A. L.:
780 Regional climate modeling for the developing world: The ICTP RegCM3 and RegCNET,

781 Bull. Am. Meteorol. Soc., 88(9), 1395–1409, doi:10.1175/BAMS-88-9-1395, 2007.

782 Pan, Y., Birdsey, R. A., Fang, J., Houghton, R., Kauppi, P. E., Kurz, W. A., Phillips, O. L.,
783 Shvidenko, A., Lewis, S. L., Canadell, J. G., Ciais, P., Jackson, R. B., Pacala, S. W.,
784 McGuire, A. D., Piao, S., Rautiainen, A., Sitch, S. and Hayes, D.: A Large and Persistent
785 Carbon Sink in the World’s Forests, *Science*, 333(6045), 988–993,
786 doi:10.1126/science.1201609, 2011.

787 Pasztor, F., Matulla, C., Rammer, W. and Lexer, M. J.: Drivers of the bark beetle disturbance
788 regime in Alpine forests in Austria, *For. Ecol. Manage.*, 318, 349–358,
789 doi:10.1016/j.foreco.2014.01.044, 2014.

790 Perring, M. P., Hedin, L. O., Levin, S. A., McGroddy, M. and de Mazancourt, C.: Increased
791 plant growth from nitrogen addition should conserve phosphorus in terrestrial
792 ecosystems., *Proc. Natl. Acad. Sci. U. S. A.*, 105(6), 1971–6,
793 doi:10.1073/pnas.0711618105, 2008.

794 Pretzsch, H., Biber, P., Schütze, G., Uhl, E. and Rötzer, T.: Forest stand growth dynamics in
795 Central Europe have accelerated since 1870, *Nat. Commun.*, 5(4967),
796 doi:10.1038/ncomms5967, 2014.

797 Radu, R., Déqué, M. and Somot, S.: Spectral nudging in a spectral regional climate model,
798 *Tellus A*, 60(5), 898–910, doi:10.1111/j.1600-0870.2008.00341.x, 2008.

799 Rammer, W. and Seidl, R.: Coupling human and natural systems: Simulating adaptive
800 management agents in dynamically changing forest landscapes, *Glob. Environ. Chang.*,
801 35, 475–485, doi:10.1016/j.gloenvcha.2015.10.003, 2015.

802 R Development Core Team: R: A language and environment for statistical computing. R
803 Foundation for Statistical Computing, Vienna, Austria. <http://R-project.org>, 2017.

804 Reichstein, M., Bahn, M., Ciais, P., Frank, D., Mahecha, M. D., Seneviratne, S. I.,
805 Zscheischler, J., Beer, C., Buchmann, N., Frank, D. C., Papale, D., Rammig, A., Smith,
806 P., Thonicke, K., van der Velde, M., Vicca, S., Walz, A. and Wattenbach, M.: Climate
807 extremes and the carbon cycle, *Nature*, 500(7462), 287–295, doi:10.1038/nature12350,
808 2013.

809 Reyer, C., Lasch-Born, P., Suckow, F., Gutsch, M., Murawski, A. and Pilz, T.: Projections of
810 regional changes in forest net primary productivity for different tree species in Europe
811 driven by climate change and carbon dioxide, *Ann. For. Sci.*, 71(2), 211–225,
812 doi:10.1007/s13595-013-0306-8, 2014.

813 Roth, T., Kohli, L., Rihm, B. and Achermann, B.: Nitrogen deposition and diversity at the
814 landscape scale *Subje, R. Soc. open Sci.*, 2(150017), 1–8, 2015.

815 Schelhaas, M. J.: The wind stability of different silvicultural systems for Douglas-fir in the
816 Netherlands: A model-based approach, *Forestry*, 81(3), 399–414,
817 doi:10.1093/forestry/cpn028, 2008.

818 Schurman, J. S., Trotsiuk, V., Bače, R., Čada, V., Fraver, S., Janda, P., Kulakowski, D.,
819 Labusova, J., Mikoláš, M., Nagel, T. A., Seidl, R., Synek, M., Svobodová, K.,
820 Chaskovskyy, O., Teodosiu, M. and Svoboda, M.: Large-scale disturbance legacies and
821 the climate sensitivity of primary *Picea abies* forests, *Glob. Chang. Biol.*, 38(1), 42–49,
822 doi:10.1111/gcb.14041, 2018.

823 Schwaab, J., Bavay, M., Davin, E., Hagedorn, F., Hüsler, F., Lehning, M., Schneebeli, M.,
824 Thürig, E. and Bebi, P.: Carbon storage versus albedo change: Radiative forcing of forest
825 expansion in temperate mountainous regions of Switzerland, *Biogeosciences*, 12(2), 467–
826 487, doi:10.5194/bg-12-467-2015, 2015.

827 Seidl, R. and Rammer, W.: Climate change amplifies the interactions between wind and bark
828 beetle disturbances in forest landscapes, *Landsc. Ecol.*, 32(7) , doi:10.1007/s10980-016-
829 0396-4, doi:10.1007/s10980-016-0396-4, 2017.

830 Seidl, R., Schelhaas, M.-J., Lindner, M. and Lexer, M. J.: Modelling bark beetle disturbances
831 in a large scale forest scenario model to assess climate change impacts and evaluate
832 adaptive management strategies, *Reg. Environ. Chang.*, 9(2), 101–119,
833 doi:10.1007/s10113-008-0068-2, 2009.

834 Seidl, R., Schelhaas, M.-J. and Lexer, M. J.: Unraveling the drivers of intensifying forest
835 disturbance regimes in Europe, *Glob. Chang. Biol.*, 17(9), 2842–2852,
836 doi:10.1111/j.1365-2486.2011.02452.x, 2011.

837 Seidl, R., Rammer, W., Scheller, R. M. and Spies, T. A.: An individual-based process model
838 to simulate landscape-scale forest ecosystem dynamics, *Ecol. Modell.*, 231, 87–100,
839 doi:10.1016/j.ecolmodel.2012.02.015, 2012a.

840 Seidl, R., Spies, T. A., Rammer, W., Steel, E. A., Pabst, R. J. and Olsen, K.: Multi-scale
841 drivers of spatial variation in old-growth forest carbon density disentangled with Lidar
842 and an individual-based landscape model, *Ecosystems*, 15(8), 1321–1335,
843 doi:10.1007/s10021-012-9587-2, 2012b.

844 Seidl, R., Rammer, W. and Spies, T. A.: Disturbance legacies increase the resilience of forest
845 ecosystem structure, composition, and functioning, *Ecol. Appl.*, 24(8), 2063–2077,
846 doi:10.1890/14-0255.1, 2014a.

847 Seidl, R., Schelhaas, M.-J., Rammer, W. and Verkerk, P. J.: Increasing forest disturbances in
848 Europe and their impact on carbon storage, *Nat. Clim. Chang.*, 4(9), 806–810,
849 doi:10.1038/nclimate2318, 2014b.

850 Seidl, R., Rammer, W. and Blennow, K.: Simulating wind disturbance impacts on forest
851 landscapes: Tree-level heterogeneity matters, *Environ. Model. Softw.*, 51, 1–11,
852 doi:10.1016/j.envsoft.2013.09.018, 2014c.

853 Seidl, R., Donato, D. C., Raffa, K. F. and Turner, M. G.: Spatial variability in tree
854 regeneration after wildfire delays and dampens future bark beetle outbreaks, *Proc. Natl.*
855 *Acad. Sci.*, 113(46), 13075–13080, doi:10.1073/pnas.1615263113, 2016.

856 Seidl, R., Thom, D., Kautz, M., Martin-Benito, D., Peltoniemi, M., Vacchiano, G., Wild, J.,
857 Ascoli, D., Petr, M., Honkaniemi, J., Lexer, M. J., Trotsiuk, V., Mairota, P., Svoboda,
858 M., Fabrika, M., Nagel, T. A. and Reyer, C. P. O.: Forest disturbances under climate
859 change, *Nat. Clim. Chang.*, 7(6), 395–402, doi:10.1038/nclimate3303, 2017.

860 Senf, C. and Seidl, R.: Natural disturbances are spatially diverse but temporally synchronized
861 across temperate forest landscapes in Europe, *Glob. Chang. Biol.*, 24(3), 1201–1211,
862 doi:10.1111/gcb.13897, 2018.

863 Silva Pedro, M., Rammer, M. and Seidl, R. Tree species diversity mitigates disturbance
864 impacts on the forest carbon cycle, *Oecol.*, 177(3), 619–630, doi: 10.1007/s00442-014-
865 3150-0

866 Sitch, S., Huntingford, C., Gedney, N., Levy, P. E., Lomas, M., Piao, S. L., Betts, R., Ciais,
867 P., Cox, P., Friedlingstein, P., Jones, C. D., Prentice, I. C. and Woodward, F. I.:
868 Evaluation of the terrestrial carbon cycle, future plant geography and climate-carbon
869 cycle feedbacks using five Dynamic Global Vegetation Models (DGVMs), *Glob. Chang.*
870 *Biol.*, 14(9), 2015–2039, doi:10.1111/j.1365-2486.2008.01626.x, 2008.

871 Soyka, W.: Die Borkenkäferverheerungen in Reichraming und ihre Bekämpfung, *Allg.*
872 *Forst- und Jagdzeitung*, 54, 155–156, 1936.

873 Stadelmann, G., Bugmann, H., Wermelinger, B., Meier, F. and Bigler, C.: A predictive
874 framework to assess spatio-temporal variability of infestations by the european spruce
875 bark beetle, *Ecography*, 36(11), 1208–1217, doi:10.1111/j.1600-0587.2013.00177.x,
876 2013.

877 Steffen, W., Persson, Å., Deutsch, L., Zalasiewicz, J., Williams, M., Richardson, K., Crumley,
878 C., Crutzen, P., Folke, C., Gordon, L., Molina, M., Ramanathan, V., Rockström, J.,
879 Scheffer, M., Schellnhuber, H. J. and Svedin, U.: The anthropocene: From global change
880 to planetary stewardship, *Ambio*, 40(7), 739–761, doi:10.1007/s13280-011-0185-x,
881 2011.

882 Svoboda, M., Janda, P., Nagel, T. a., Fraver, S., Rejzek, J. and Bače, R.: Disturbance history
883 of an old-growth sub-alpine *Picea abies* stand in the Bohemian Forest, Czech Republic, *J.*
884 *Veg. Sci.*, 23(1), 86–97, doi:10.1111/j.1654-1103.2011.01329.x, 2012.

885 Temperli, C., Zell, J., Bugmann, H. and Elkin, C.: Sensitivity of ecosystem goods and
886 services projections of a forest landscape model to initialization data, *Landsc. Ecol.*,
887 28(7), 1337–1352, doi:10.1007/s10980-013-9882-0, 2013.

888 Thom, D., Seidl, R., Steyrer, G., Krehan, H. and Formayer, H.: Slow and fast drivers of the
889 natural disturbance regime in Central European forest ecosystems, *For. Ecol. Manage.*,
890 307, 293–302, doi:10.1016/j.foreco.2013.07.017, 2013.

891 Thom, D., Rammer, W. and Seidl, R.: Disturbances catalyze the adaptation of forest
892 ecosystems to changing climate conditions, *Glob. Chang. Biol.*, 23(1), 269–282,
893 doi:10.1111/gcb.13506, 2017c.

894 Thom, D., Rammer, W. and Seidl, R.: The impact of future forest dynamics on climate:
895 interactive effects of changing vegetation and disturbance regimes, *Ecol. Monogr.*, 87(4),

896 665–684, doi:10.1002/ecm.1272, 2017a.

897 Thom, D., Rammer, W., Dirnböck, T., Müller, J., Kobler, J., Katzensteiner, K., Helm, N. and
898 Seidl, R.: The impacts of climate change and disturbance on spatio-temporal trajectories
899 of biodiversity in a temperate forest landscape, *J. Appl. Ecol.*, 54(1), 28–38,
900 doi:10.1111/1365-2664.12644, 2017b.

901 Thornton, P. E. and Rosenbloom, N. A.: Ecosystem model spin-up: Estimating steady state
902 conditions in a coupled terrestrial carbon and nitrogen cycle model, *Ecol. Modell.*,
903 189(1–2), 25–48, doi:10.1016/j.ecolmodel.2005.04.008, 2005.

904 Valinger, E. and Fridman, J.: Factors affecting the probability of windthrow at stand level as a
905 result of Gudrun winter storm in southern Sweden, *For. Ecol. Manage.*, 262(3), 398–403,
906 doi:10.1016/j.foreco.2011.04.004, 2011.

907 Weichenberger, J.: Die Holztrift im Nationalpark Kalkalpen Teil 1: Bestandsaufnahme,
908 Leonstein., 1994.

909 Weichenberger, J.: Die Holztrift im Nationalpark Kalkalpen Teil 2: Geschichtliche
910 Aufarbeitung, Leonstein., 1995.

911 Weinfurter, P.: 80 Jahre Bundesforste. Geschichte der Österreichischen Bundesforste,
912 Purkersdorf., 2005.

913 Wingfield, M. J., Barnes, I., de Beer, Z. W., Roux, J., Wingfield, B. D. and Taerum, S. J.:
914 Novel associations between ophiostomatoid fungi, insects and tree hosts: current status—
915 future prospects, *Biol. Invasions*, 19(11), 3215–3228, doi:10.1007/s10530-017-1468-3,
916 2017.

917 Wu, H.-I., Sharpe, P. J. H., Walker, J. and Penridge, L. K.: Ecological field theory: A spatial
918 analysis of resource interference among plants, *Ecol. Modell.*, 29, 215–243, 1985.

920 **Tables**

921 Table 1. Development of total ecosystem carbon stocks (tC ha⁻¹) over time and in different scenarios of disturbance and land use history as well as
 922 future climate. Values are based on iLand simulations and indicate means and standard deviations (SD) over averaged landscape values of the
 923 replicates in the respective scenarios. “Historic climate” assumes the continuation of the climate 1950 – 2010 throughout the 21st century, while
 924 “Climate change” summarizes the effect of three alternative climate change scenarios for the 21st century. The first three columns indicate the
 925 respective permutation of the simulated disturbance and land use history, with the first line representing the historical reconstruction of landscape
 926 development. Y=yes, N=no.

First nat. dist. episode	Land use	Second nat. dist. episode	year 1905		year 1923		year 1997		year 2013		Historic climate year 2099		Climate change year 2099	
			mean	SD	mean	SD	mean	SD	mean	SD	mean	SD	mean	SD
Y	Y	Y	303.5	<0.1	331.1	<0.1	403.2	0.7	427.8	0.8	487.7	0.7	466.4	23.7
Y	N	Y	303.5	<0.1	331.2	<0.1	457.5	0.6	466.7	0.7	487.2	1.0	463.3	20.9
Y	Y	N	303.5	<0.1	331.0	<0.1	403.2	0.7	430.6	0.7	488.2	0.7	467.0	23.3
Y	N	N	303.5	<0.1	331.2	<0.1	457.5	0.5	470.9	0.7	487.3	0.7	463.4	21.1
N	Y	Y	303.5	0.1	332.7	0.1	404.3	0.8	428.8	0.8	487.8	0.8	466.3	23.7
N	N	Y	303.5	0.1	333.0	0.1	458.7	0.5	468.0	0.6	487.8	0.8	464.0	21.3
N	Y	N	303.5	0.1	332.7	0.1	404.2	0.7	431.3	0.8	488.3	0.9	466.4	23.6
N	N	N	303.5	0.1	333.0	0.1	458.6	0.5	471.7	0.6	487.9	0.9	464.1	21.0

927

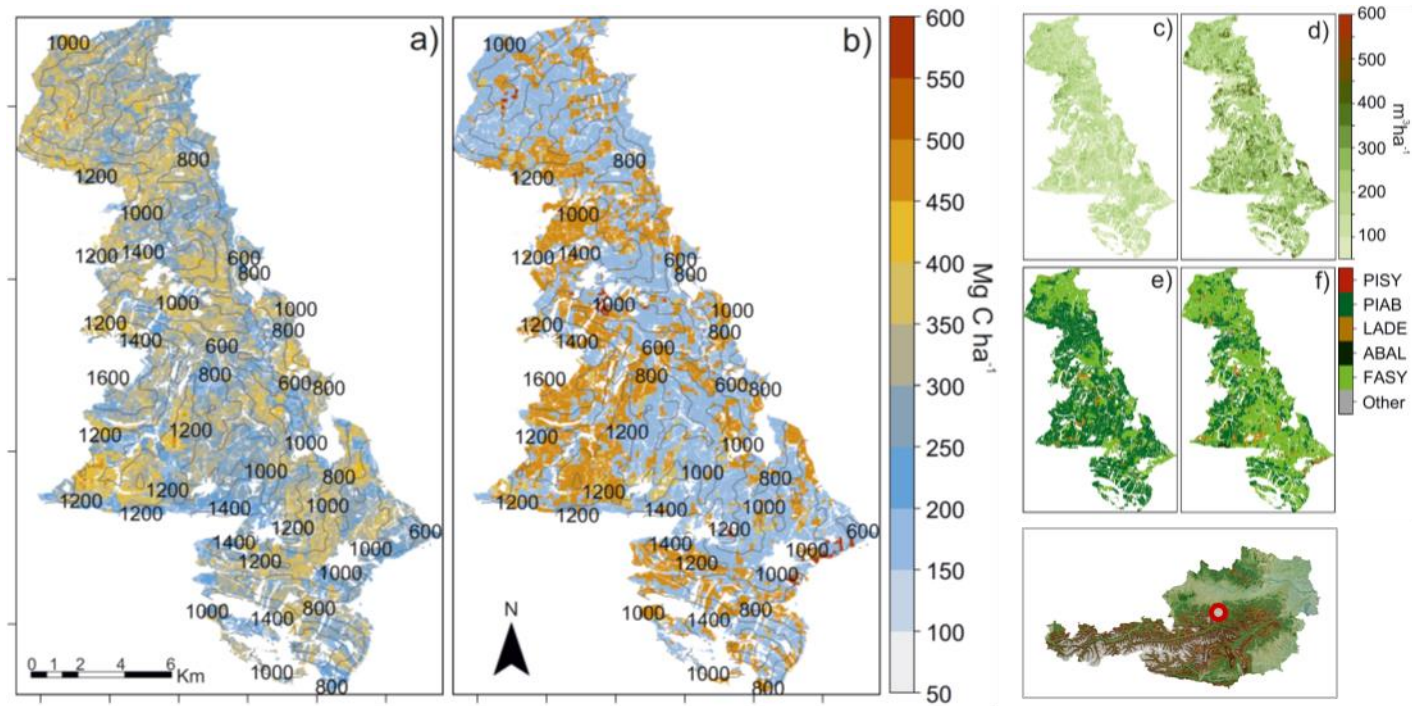
928

929 Table 2. Growing stock by tree species (m³ ha⁻¹). Values are based on iLand simulation runs and indicate species means and standard deviation (SD)
 930 over averaged landscape values of the replicates in the respective scenarios. “Historic climate” assumes the continuation of the climate 1950 – 2010
 931 throughout the 21st century, while “Climate change” summarizes the effect of three alternative climate change scenarios for the 21st century.

Tree species	year 1905		year 1923		year 1997		year 2013		Historic climate year 2099		Climate change year 2099	
	mean	SD	mean	SD	mean	SD	mean	SD	mean	SD	mean	SD
<i>Abies alba</i>	4.2	0.0	2.1	0.0	9.7	2.2	12.7	2.6	28.7	6.1	33.7	7.6
<i>Fagus sylvatica</i>	68.0	0.6	76.8	0.6	165.6	39.8	198.5	34.4	286.8	2.8	309.7	19.7
<i>Larix decidua</i>	21.5	0.2	23.9	0.2	41.7	5.2	40.5	9.7	17.4	7.9	16.2	7.1
<i>Picea abies</i>	116.3	0.5	138.6	0.5	235.7	43.6	250.8	40.5	276.3	36.6	229.9	33.6
Other tree species	2.3	0.2	6.0	0.2	14.7	1.4	16.0	1.6	13.4	0.5	23.8	1.7
Total	212.3	0.8	247.4	0.8	467.4	79.0	518.5	66.0	622.6	35.4	613.3	46.5

932

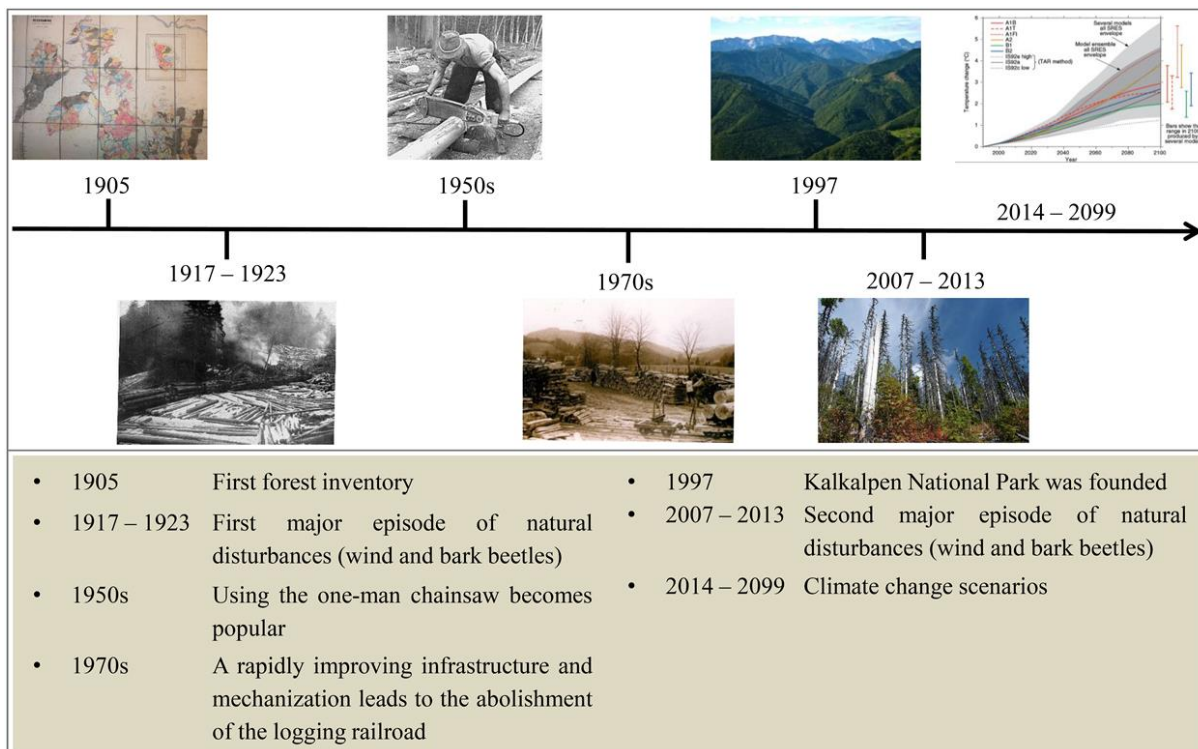
933 **Figures**



934

935 Fig. 1: State of forest ecosystem attributes across the study landscape in 1905 and 2013 as well as location of the landscape in Austria (lower right
 936 panel). Panels (a) and (b) show the distribution of total ecosystem carbon, while panels (c) and (d) present growing stock, and panels (e) and (f)
 937 indicate the dominant tree species (i.e., the species with the highest growing stock in a 100m pixel) in 1905 and 2013, respectively. PISY = *Pinus*
 938 *syvestris*, PIAB = *Picea abies*, LADE = *Larix decidua*, ABAL = *Abies alba*, FASY = *Fagus sylvatica*. “Other” refers to either other dominant species

939 not individually listed here due to their low abundance, or areas where no trees are present. Isolines represent elevational gradients in the landscape
940 (in m asl).



941

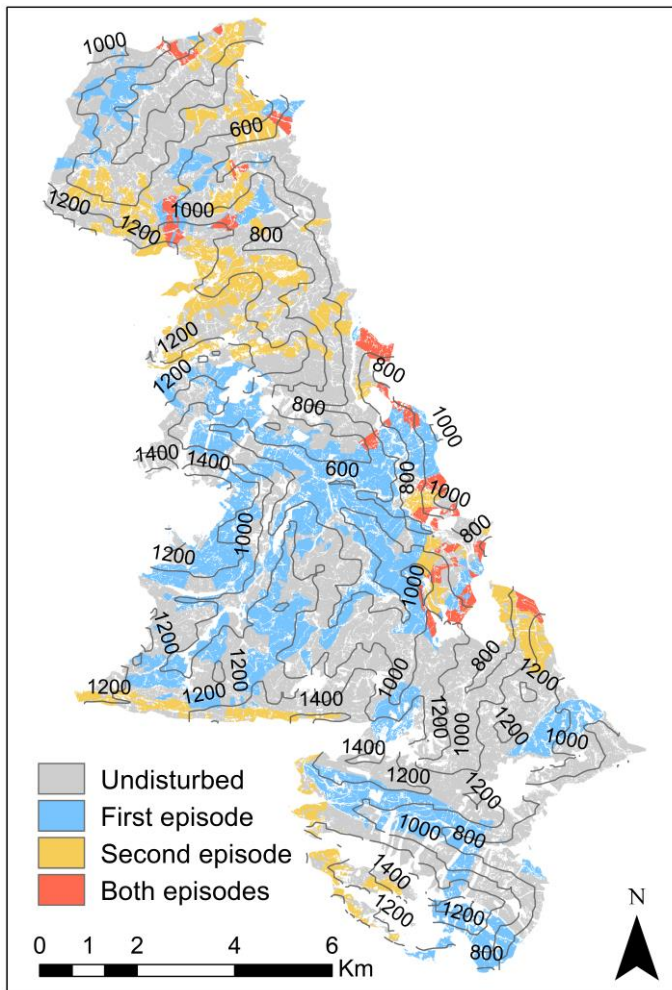
942 Fig. 2. Timeline of historic events of relevance for the simulation of the study landscape. Image

943 credits: 1905 and 1917 – 1923: archives of the Austrian Federal Forests; 1950s:

944 <https://waldwissen.at>; 1970s: <https://atterwiki.at>; 1997: <http://kalkalpen.at>; 2007 – 2013: photo

945 taken by the authors of this study; 2014 – 2099: <http://climate-scenarios.canada.ca>.

946

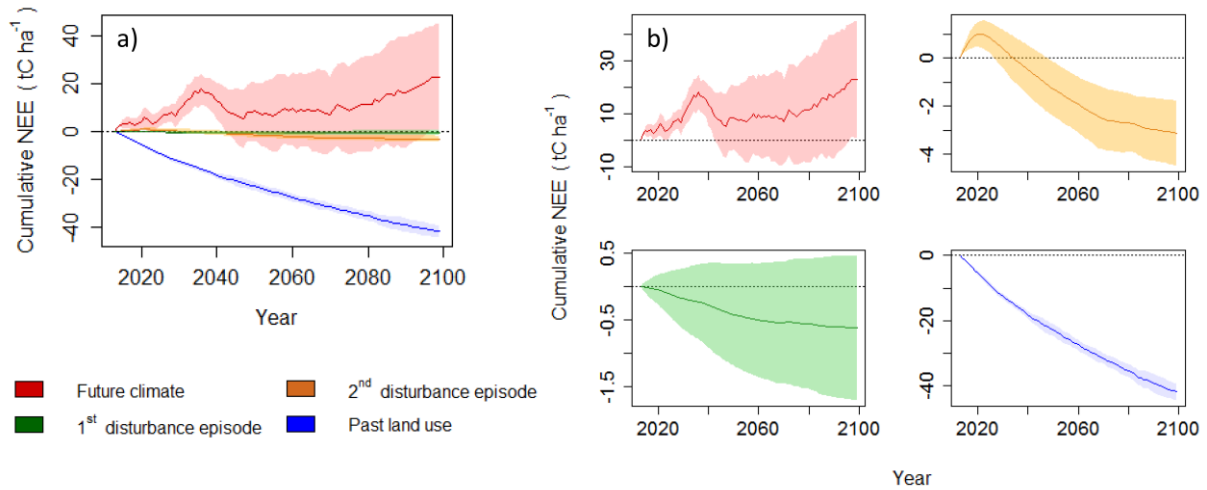


948

949 Fig. 3: Disturbance activity in two episodes of natural disturbance, from 1917 – 1923 (first
 950 episode) and 2007 – 2013 (second episode). Isolines represent elevational gradients (in m asl).

951

952



953

954 Fig. 4. Scenarios of mean cumulative change in future net ecosystem exchange (NEE). Panel
955 (a) shows the effects of all considered drivers of NEE change on the same scale while panel (b)
956 zooms into the individual effect of each driver. Cumulative NEE was analyzed after the second
957 disturbance episode (setting NEE to 0 in year 2013) to allow for the simultaneous representation
958 of the long-term legacy effects of different past disturbance events (i.e., the first (1917-1923)
959 and second (2007-2013) disturbance episode, respectively) and land use change (i.e.,
960 management ceased in 1997) as well as future climate change. Differences in NEE were
961 derived from a factorial simulation experiment, comparing each factor to its baseline (e.g.,
962 future climate scenarios to baseline climate) while keeping all other factors constant. Shaded
963 areas denote the standard deviation in NEE for the respective scenarios. NEE is the carbon flux
964 from the ecosystem to the atmosphere (i.e., $\text{NEE} = -\text{NEP}$). Note that y-axis scales differ for
965 each panel.