



1 **Disturbance legacies have a stronger effect on future carbon exchange than**
2 **climate in a temperate forest landscape**

3

4 **Running head:** “Disturbance legacies determine C exchange”

5

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14 **Abstract**

15 Forest ecosystems play an important role in the global climate system, and are thus intensively
16 discussed in the context of climate change mitigation. Over the past decades temperate forests
17 were a carbon (C) sink to the atmosphere. However, it remains unclear to which degree this C
18 uptake is driven by a recovery from past disturbances vs. ongoing climate warming, inducing
19 high uncertainty regarding the future temperate forest C sink. Here our objectives were (i) to
20 investigate legacies within the natural disturbance regime by empirically analyzing two
21 disturbance episodes affecting the same landscape 90 years apart, and (ii) to unravel the effects
22 of past disturbances and future climate on 21st century forest C uptake by means of simulation
23 modelling. We collected historical data from archives to reconstruct vegetation and disturbance
24 history of a forest landscape in the Austrian Alps from 1905 to 2013. The effect of past legacies
25 and future climate was determined by simulating 32 different combinations of past disturbances
26 (including natural disturbances and management) and future climate scenarios. We found only
27 moderate spatial overlap between two episodes of wind and bark beetle disturbance affecting
28 the landscape in the early 20th and 21st century, respectively. The future forest C sink was driven
29 by past disturbances, while climate change reduced forest C uptake. Historic management (and
30 its cessation) had a considerably stronger influence on the future C balance than the natural
31 disturbance episodes of the past. We conclude that neglecting disturbance legacies can
32 substantially bias assessments of future forest dynamics.

33

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

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42 **1. Introduction**

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

56 For the future, dynamic Global Vegetation Models (DGVMs) frequently suggest a persistent
57 forest carbon sink (Keenan et al., 2016; Sitch et al., 2008). However, while DGVMs are suitable
58 for tracking the direct effects of global change, they frequently neglect the effects of
59 disturbances and their long-term legacy. Both natural and anthropogenic disturbances have
60 decreased the amount of carbon currently stored in forest ecosystems (Erb et al., 2018; Goetz



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

77 Here we investigate the effect of long-term disturbance legacies on forest ecosystem dynamics,
78 in order to better understand the drivers of future forest carbon uptake, and thus aid the
79 development of effective climate change mitigation strategies. In particular, our first objective
80 was to empirically investigate the temporal interaction of two major episodes of natural
81 disturbance affecting the same Central European forest landscape 90 years apart (i.e., 1917 –
82 1923 and 2007 – 2013). We hypothesized a temporal autocorrelation of the two major
83 disturbance episodes, and specifically an amplifying effect from the earlier disturbance episode
84 on the later disturbance episode, based on recent observations of centennial disturbance waves
85 in Europe's forests (Schurman et al., 2018). Our second goal was to quantify the contribution



86 of past disturbances (both natural and anthropogenic) on the future C uptake of the landscape
87 under a number of climate change scenarios using simulation modelling. We were particularly
88 interested in the relative effects of past disturbances and future climate scenarios on the future
89 forest C sink strength. To that end we reconstructed the vegetation and disturbance history of
90 the landscape from 1905 to 2013 using historical sources and remote sensing. We subsequently
91 determined the effect of past disturbances on 21st century C dynamics by simulating forests
92 from the early 20th century to the end of the 21st century, experimentally altering past
93 disturbance regimes in a factorial simulation experiment. These analyses were run under
94 multiple climate scenarios for the 21st century, and focused on Net Ecosystem Exchange (NEE)
95 (i.e., the net C exchange of the ecosystem with the atmosphere) as the response variable. We
96 hypothesized that the legacy of past disturbances (management + natural causes) is of
97 paramount importance for the future carbon sink (Thom et al., 2017a), expecting a saturation
98 of carbon uptake as the landscape recovers from past disturbances (i.e., a negative but
99 decreasing NEE through the 21st century). Moreover, we hypothesized a negative impact of
100 future climate change on carbon uptake as a result of less favorable conditions for carbon-rich
101 spruce dominated forests (Thom et al., 2017a).

102

103 **2. Materials and Methods**

104 **2.1 Study area**

105 We selected a 7,609 ha forest landscape located in the northern front range of the Alps as our
106 study area (Fig. 1). Focusing on the landscape scale allowed us to mechanistically capture
107 changes in forest structure and C stocks by jointly considering processes at the large scale such
108 as disturbances as well as fine scale processes such as competition between individual trees.
109 The focal landscape is particularly suited to address our research questions as it (i) was affected



110 by two major episodes of natural disturbance (driven by wind and bark beetles) in the past
111 century, and (ii) has a varied management history, with intensive management up until 1997,
112 and then becoming a part of Kalkalpen National Park (KANP), the largest contiguous protected
113 forest area in Austria. The steep elevational gradient of the study landscape, ranging from 414
114 m to 1637 m a.s.l., results in large variation in environmental conditions. For instance,
115 temperatures range from 4.3 – 9.0°C and mean annual precipitation sums vary between 1179 –
116 1648 mm on the landscape. Shallow Lithic and Renzic Leptosols as well as Chromic Cambisols
117 over calcareous bedrock are the prevailing soil types (Kobler 2004). The most prominent natural
118 forest types on the landscape are European beech (*Fagus sylvatica* [L.]) dominated forests at
119 low elevations, mixtures of Norway spruce (*Picea abies* [K.]), silver fir (*Abies alba* [Mill.]) and
120 European beech at mid-elevations, and Norway spruce dominated forests at high elevations.
121 These forest types are among the most common ones in Europe, and are highly valuable to
122 society also from a socio-economic perspective (Hanewinkel et al., 2012).

123

124 **2.2 Simulation model**

125 We employed the individual-based forest landscape and disturbance model (iLand) to simulate
126 past and future forest dynamics at our study landscape. iLand is a high-resolution process-based
127 forest model, designed to simulate the dynamic feedbacks between vegetation, climate and
128 disturbance regimes (Seidl et al., 2012a, 2012b). It simulates processes in a hierarchical multi-
129 scale framework, i.e., considering processes at the individual tree (e.g., growth, mortality as
130 well as competition for light, water, and nutrients), stand (e.g., water and nutrient availability),
131 and landscape (e.g., seed dispersal, disturbances) scale as well as their cross-scale interactions.
132 Competition for resources among individual trees is based on ecological field theory (Wu et al.,
133 1985). Resource utilization is modelled employing a light use efficiency approach (Landsberg



134 and Waring, 1997), incorporating the effects of temperature, solar radiation, vapor pressure
135 deficit, soil water and nutrient availability on a daily basis. Resource use efficiency is further
136 modified by variation in the atmospheric CO₂ concentration. Seeds are dispersed via species-
137 specific dispersal kernels (20 × 20 m horizontal resolution) around individual mature trees. The
138 establishment success of the regeneration is constrained by environmental filters (e.g.,
139 temperature and light availability).

140 Mortality of trees is driven by stress-induced carbon starvation and also considers a stochastic
141 probability of tree death depending on life-history traits. Additionally, iLand includes three
142 submodules to simulate natural disturbances, including wind (Seidl et al., 2014c), bark beetles
143 (Seidl and Rammer 2017), and wildfire (Seidl et al., 2014b). As wind and bark beetles are of
144 paramount importance for the past and future disturbance regimes of Central Europe's forests
145 (Seidl et al., 2014a; Thom et al., 2013), we employed only these two process-based disturbance
146 submodules in our simulations. The impact of wind disturbance in iLand depends on species-
147 and size-specific susceptibility (e.g., critical wind speeds of uprooting and stem breakage),
148 vertical forest structure (e.g., gaps), and storm characteristics (e.g., maximum wind speeds).
149 The bark beetle module simulates the impact of *Ips typographus* (L.) on Norway spruce, and
150 thus addresses the effects of the most important bark beetle species in Europe with respect to
151 area affected and timber volume disturbed (Kautz et al., 2017; Seidl et al., 2009). The model
152 *inter alia* accounts for insect abundance, phenology and development, as well as emergence
153 and dispersal. It computes the number of beetle generations and sister broods developed per
154 year based on the prevailing climate, and considers individual tree defense capacity and
155 susceptibility. Interactions between wind and bark beetles arise from a high infestation
156 probability and low defense capacity of freshly downed trees after wind disturbance, while
157 newly formed gaps (e.g., by bark beetles) increase the exposure of surrounding forests to storm
158 events.



159 In addition to the submodules of natural disturbance we used the agent-based forest
160 management module (ABE) in iLand (Rammer and Seidl, 2015) to simulate past forest
161 disturbances by management. ABE enables the dynamic application of generalized stand
162 treatment programs, including planting, tending, thinning, and harvesting activities. The
163 dynamically simulated management agent observes constraints at the stand and landscape
164 scales, such as maximum clearing sizes and sustainable harvest levels. Besides silvicultural
165 treatments, we used ABE to emulate the past management practice of salvage logging after bark
166 beetle outbreaks. A detailed description of the implementation of historic management
167 activities in the simulations can be found in the Supplementary Material (S4).

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

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

182



183 **2.3 Reconstructing forest management and disturbance history**

184 The study area has a long history of intensive timber harvesting for charcoal production, mainly
185 driven by a local pre-industrial iron-producing syndicate. This syndicate was active until 1889,
186 when the land was purchased by the k.k. (“kaiserlich und königlich”) Ministry for Agriculture.
187 During the 20th century, the majority of the landscape was managed by the Austrian Federal
188 Forests, and only limited areas within the landscape were still under the ownership of industrial
189 private companies (Weichenberger, 1994, 1995; Weinfurter, 2005). Forest management in the
190 late 19th and early 20th century was strongly influenced by the emerging industrialization. The
191 substitution of wood by mineral coal for heating, but especially for industrial energy supply,
192 changed the focus of forest management from fuel wood to timber production. At the same
193 time, an increase in agricultural productivity (also triggered by input of fossil resources as well
194 as artificial fertilizer) allowed for the abandonment of less productive agricultural plots, often
195 followed by afforestation or natural regrowth of forest vegetation. Consequently, growing
196 stocks increased in many parts of Europe throughout the 20th century as the result of increases
197 in both forest extent and density (Bebi et al., 2017). In our study system, the shifting focus from
198 fuel wood to timber production around 1900 was accompanied by the introduction of systematic
199 stand delineation for spatial management planning (Fig. S2) and decadal inventories and forest
200 plan revisions. These documents are preserved in the archives of the Austrian Federal Forests,
201 and were used here to reconstruct past forest vegetation as well as management and disturbance
202 history (see S1, Fig. S2 and S3 in the Supplementary Material for details).

203 The oldest historic vegetation data available for the landscape were from an inventory
204 conducted between the years 1898 and 1911 and comprised growing stock and age classes for
205 11 tree species at the level of stand compartments for the entire landscape; we subsequently
206 used the year 1905 (representing the area-weighted mean year of this initial inventory) as the
207 temporal starting point for our analyses (Fig. 2). A major challenge for managers was to extract



208 resources from remote and inaccessible parts of the topographically highly complex landscape.
209 The most important means of timber transportation was drifting (i.e., flushing logs down creeks
210 and streams after artificially damming them). However, this transportation technique was not
211 feasible for heavy hardwood timber such as beech (Grabner et al., 2004). Consequently,
212 managers harvested trees selectively, and mainly focused on accessible areas (i.e., stands close
213 to streams), leading to a bimodal age distribution on the landscape in 1905 with many young
214 and several old stands (Fig. S8).

215 In addition to deriving the state of the forest in 1905, we reconstructed management activities
216 (thinnings, final harvests, artificial regeneration) and natural disturbances (wind and bark
217 beetles) until 2013. From 1905 to 1917 timber extraction was fairly low. Between 1917 and
218 1923, however, a major disturbance episode by wind and bark beetles hit the region. Resulting
219 from a lack of labor force (military draft, malnutrition) in the last year of World War I a major
220 windthrow in 1917 could not be cleared, and the resulting bark beetle outbreak affected large
221 parts of the landscape. Overall, wind and bark beetles disturbed approximately one million
222 cubic meters of timber in our study area between 1917 and 1923 (calculation from archival
223 sources; Soyka, 1936; Weichenberger, 1994). Consequently, a railroad was installed to access
224 and salvage the disturbed timber. After the containment of the disturbance in 1923 forest
225 management resumed at low intensity and no major natural disturbances were recorded.
226 Following World War II, a network of forest roads was built in order to gradually replace
227 transportation by railroads. The introduction of motorized chain saws (Fig. 2) further
228 contributed to an intensification of harvests. By 1971, forest railroads were completely replaced
229 by motorized transportation on forest roads, resulting in a further increase in the timber
230 extracted from the landscape (Fig. S9). Timber removals from management as well as natural
231 disturbances from wind and bark beetles between 1905 and 1997 were reconstructed from
232 yearly management reviews available from archival sources. With the landscape becoming part



233 of KANP forest management ceased in 1997. A second major episode of natural disturbances
234 affected the landscape from 2007-2013, when a large bark beetle outbreak followed three storm
235 events in 2007 and 2008. This second disturbance episode was reconstructed from disturbance
236 records of KANP in combination with remote sensing data (Seidl and Rammer, 2016; Thom et
237 al., 2017b).

238

239 **2.4 Landscape initialization and drivers**

240 The vegetation data for the year 1905 were derived from historical records for 2079 stands with
241 a median stand area of 1.7 ha. On average over the landscape, the growing stock was 212.3 m³
242 ha⁻¹ in 1905. The most common species were Norway spruce (with a growing stock of on
243 average 116.3 m³ ha⁻¹), European beech (68.0 m³ ha⁻¹), and European larch (*Larix decidua*
244 [Mill.], 21.5 m³ ha⁻¹). With an average growing stock of 4.2 m³ ha⁻¹ silver fir was considerably
245 underrepresented on the landscape relative to the potential natural vegetation composition,
246 resulting from historic clear-cut management and high browsing pressure from deer (see also
247 Kučeravá and others 2012). Despite these detailed data on past vegetation not all information
248 for initializing iLand were available from archival sources, e.g., diameters at breast height (dbh)
249 and height of individual trees, as well as tree positions, regeneration and belowground carbon-
250 pools had to be reconstructed by other means. To that end we developed a new method for
251 initializing vegetation in iLand, combining spin-up simulations with empirical reference data
252 on vegetation state, henceforth referred to as legacy spin-up.

253 Commonly, spin-ups run models for a certain amount of time or until specified stopping criteria
254 are reached (e.g., steady-state conditions). The actual model-based analysis is then started from
255 the thus spun-up vegetation condition (Thornton and Rosenbloom, 2005). This has the
256 advantage that the model-internal dynamics (e.g., the relationships between the different C and



257 N pools in an ecosystem) are consistent when the focal analysis starts. However, the thus
258 derived initial vegetation condition does frequently not correspond well with the vegetation
259 state observed at a given point in time, and does not account for the legacies of past management
260 and disturbance. The legacy spin-up approach developed here aims to reconstruct a (partially)
261 known reference state of the vegetation (e.g., the species composition, age, and growing stock
262 reconstructed from archival sources for the current analysis) from simulations (Fig. S5). To this
263 end iLand simulates long-term forest development for each stand, employing an approximation
264 of the past management and disturbance regime. During the simulations, the emerging forest
265 trajectory is periodically compared to the respective reference values, and the assumed past
266 management is adapted iteratively in order to decrease the difference between simulated
267 vegetation states and reference values. This procedure is executed in parallel for all stands on
268 the landscape over a long period of time (here: 1000 years), and the simulated vegetation states
269 best corresponding to the reference values are stored (including individual tree properties,
270 regeneration, and carbon pools), and later used as initial values for model-based scenario
271 analyses. A detailed description of the legacy spin-up approach is given in the Supplementary
272 Material S4.

273 In simulating 20th century forest dynamics we accounted for the abandonment of cattle grazing
274 and litter raking in forests (Glatzel, 1991) as well as an increasing deposition of nitrogen from
275 the atmosphere (Dirnböck et al., 2014; Roth et al., 2015). Specifically, we dynamically
276 modified the annual plant available nitrogen in our simulations based on data of nitrogen
277 deposition in Austria between 1880 and 2010, with nitrogen input culminating in the mid 1980s,
278 followed by a decrease and a stabilization after 2000 (Dirnböck et al., 2017). Besides edaphic
279 factors also an increase in temperature has led to more favorable conditions of tree growth
280 (Pretzsch et al., 2014). Detailed observations of climate for our study region reach back to 1950
281 (Thom et al., 2017b), requiring an extension of the climate time series to 1905. We extracted



282 data from the nearest weather station covering the period from 1905 to present (i.e., Admont,
283 located approximately 20 km south of our study area), and used its temperature and
284 precipitation record to sample years with corresponding conditions from the observational
285 record for our study landscape.

286 Simulations were run from 1905 until 2099, considering four different climate scenarios for the
287 period 2013 – 2099. Climate change was represented by three combinations of global
288 circulation models (GCM) and regional climate models (RCM) under A1B forcing, including
289 CNRM-RM4.5 (Radu et al., 2008) driven by the GCM ARPEGE, and MPI-REMO (Jacob,
290 2001), as well as ICTP-RegCM3 (Pal et al., 2007), both driven by the GCM ECHAM5. The
291 A1B scenario family assumes rapid economic growth with a global population peaking in mid-
292 century and declining thereafter, and a balanced mix of energy sources being used (IPCC 2000).
293 With average temperature increases of between +3.1°C and +3.3°C and changing annual
294 precipitation sums of -87.0 mm to +135.6 mm by the end of the 21st century, the scenarios
295 studied here are comparable to the changes expected under the representative concentration
296 pathways RCP4.5 and RCP6.0 for our study region (Thom et al., 2017c). In addition to the three
297 scenarios of climate change a historic climate scenario was simulated. The years 1950 – 2010
298 were used to represent this climatic baseline, and were randomly resampled to derive a
299 stationary climate time series until 2099.

300

301 **2.5 Analyses**

302 To address our first objective and investigate the spatio-temporal interactions of natural
303 disturbances we used the stand-level records of the two historic disturbance episodes (1917 –
304 1923 and 2007 – 2013). First, we discretized the information (disturbed/ undisturbed) and
305 rasterized the stand polygon data to a grid of 10 × 10 m. Subsequently, we used this grid to



306 calculate an odds ratio for the probability that the two disturbance events affected the same
307 locations on the landscape (i.e., the odds that areas disturbed in the first episode were disturbed
308 again in the second episode). We calculated the 95% confidence interval of the odds ratio using
309 the `vcd` package in R (Meyer et al., 2016).

310 To address our second objective and evaluate the impact of past disturbances and future climate
311 on the 21st century carbon sink strength, we ran simulations under a combination of different
312 disturbance histories and climate futures. Specifically, we experimentally permuted
313 disturbances between 1905 and 2013, and analyzed the effect of these permutations by
314 continuing the simulations until the end of the 21st century. At three points in time a bifurcation
315 of the disturbance history was considered in the simulation, resulting in eight different pathways
316 of past landscape dynamics. The three bifurcations were (i) the inclusion or omission of the first
317 episode of natural disturbance (1917-1923), (ii) a continuation of management until the
318 founding of the national park 1997 or a cessation of forest management after 1923, and (iii) the
319 inclusion or omission of the second natural disturbance episode (2007-2013) (Fig. 3). This
320 factorial permutation of elements of the actual disturbance history of the landscape was chosen
321 to assess the effects of both past and recent episodes of natural disturbance on future C uptake,
322 as well as to quantify the role of past management, while accounting for the dynamic
323 interactions between these factors in the simulation (e.g., between first and second episode of
324 natural disturbance). After 2013 four different climate scenarios were simulated for all
325 alternative disturbance histories, to assess the impacts of climate change on the future NEE of
326 the landscape.

327 All simulations were started from the landscape conditions in 1905, determined by means of
328 the legacy spin-up procedure described above. From 1905 to 1923 management and natural
329 disturbances were implemented in the simulation as recorded in the stand-level archival
330 sources. After 1923, natural disturbances were simulated dynamically using the respective



331 iLand disturbance modules. For the second disturbance episode (2007 – 2013) the observed
332 peak wind speeds for the storms Kyrill (2007), Emma (2008) and Paula (2008) were used in the
333 simulation (see Seidl and Rammer 2017 for details). Beyond 2013, natural disturbances were
334 dynamically simulated with iLand. We randomly sampled annual peak wind speeds from the
335 distribution of years 1924 – 2006 and simulated the wind and bark beetle dynamics emerging
336 on the landscape (see also Thom et al., 2017a).

337 Management interventions from 1923 to 1997 were simulated using ABE. The individual
338 silvicultural decisions were thus implemented dynamically by the management agent in the
339 model, based on generic stand treatment programs of past management in Austria's federal
340 forests and the emerging state of the forest. The advantage of this approach was that
341 management was realistically adapted to different forest states in the simulations, e.g., with
342 harvesting patterns differing in the runs in which the disturbance episode 1917 – 1923 was
343 omitted. Moreover, in line with the technical revolutions of the 20th century (Fig. 2) the
344 simulated management agent was set to account for an intensification of forest management
345 over time (e.g., a higher number of thinnings and shorter rotation periods). In summary, our
346 simulation design consisted of 32 combinations of different disturbance histories and climate
347 futures, which were replicated 20 times (i.e., in total 640 simulation runs) for the years 1905 –
348 2099 (195 years).

349 We evaluated the ability of iLand to reproduce past human and natural disturbances as well as
350 the resultant forest vegetation dynamics on the landscape by comparing simulations of the
351 baseline scenario (i.e., including historic climate, as well as reconstructed natural disturbances
352 and forest management) with independent empirical data for different time periods: The
353 simulated amount of timber extracted was compared to historical records for three time periods
354 divided by major technical revolutions during the 20th century (Fig. 2). Simulated impacts of
355 the second disturbance episode (2007 – 2013) on growing stock were compared against



356 empirical records from KANP. Simulated species shares and total growing stock were
357 compared against independent data for the year 1905, testing the ability of the legacy spin-up
358 to recreate the initial vegetation state. Furthermore, simulated species shares and growing stocks
359 were also related to observations for 1999, i.e., testing the capacity of iLand to faithfully
360 reproduce forest conditions after 95 years of vegetation dynamics. The results of all these tests
361 can be found in the Supplement of this study.

362 We used simulation outputs to investigate the changes in NEE over time and to compare the
363 different scenarios. NEE denotes the net C flux from the ecosystem to the atmosphere, with
364 negative values indicating ecosystem C gain (Chapin et al., 2006). To determine the impact of
365 past disturbances and future climate on the 21st century carbon balance of the landscape, we
366 first computed the cumulative NEE over the period 2014 – 2099 for each simulation. Next, the
367 effects of past disturbances and future climate were calculated from mean differences between
368 the different factor combinations of the simulation experiment with regard to their cumulative
369 NEE in 2099. P-values were computed by means of permutation-based independence tests
370 using the coin package (Hothorn and others 2017), and subsequently transformed into
371 confidence intervals for visualization (Altman 2011). All analyses were performed using the R
372 language and environment for statistical computing (R Development Core Team 2017).

373

374 **3. Results**

375 **3.1 Reconstructing historic landscape dynamics**

376 Using iLand, we were able to successfully reproduce historic vegetation and disturbance
377 dynamics on the landscape. The results from the legacy spin-up revealed a good match with the
378 species composition and growing stock expected from the historic records for the year 1905



379 (see S4, Fig. S6, Fig. S7). Furthermore, the iLand management module ABE was well able to
380 reproduce the intensification of forest management over the 20th century (Fig. S9). Only the
381 first evaluation period (1924 – 1952) resulted in a small overestimation of simulated harvests.
382 Further, the simulated wind and bark beetle disturbances between 2007 and 2013 corresponded
383 well to the expected values derived from KANP inventories (Fig. S10). Our dynamic simulation
384 approach adequately reproduced the tree species composition and growing stock at the
385 landscape scale after 95 years of simulation (Fig. S11). Despite an intensification of harvests
386 until 1997 and the occurrence of a major disturbance event in 1917 – 1923, the average growing
387 stock on the landscape doubled between 1905 and 2013 (Fig. S12). At the same time total
388 ecosystem carbon increased by 40.9% (Fig. S13). European beech dominance increased over
389 the 20th century, in particular at lower elevations (Fig. S12, Fig. 1e and 1f). Further details on
390 historic landscape development can be found in the Supplement S4 and Fig. S5-S13.

391

392 **3.2 Long-term temporal interactions of natural disturbances**

393 We used the empirically derived spatial footprint of two episodes of natural disturbance 90
394 years apart to investigate the long-term temporal interactions between disturbances. Both
395 disturbance episodes were found to have a similar impact on growing stock (117,441 m³ and
396 93,084 m³ of growing stock disturbed at the landscape, respectively), whereas the first episode
397 affected more than twice the area of the second episode (2334 ha and 1116 ha, respectively).
398 Only 9.2% of the area disturbed during the first episode was also affected by the second episode
399 (Fig. 4). Whereas the first disturbance episode mainly affected the central and southern reaches
400 of the study area, the effects of the second disturbance episode were most pronounced in the
401 northern parts of the landscape. The odds ratio of 0.49 ($p < 0.001$) revealed a lower probability
402 that the same location of the first disturbance episode is affected by the second disturbance



403 episode on the landscape compared to the odds that a previously undisturbed area is disturbed
404 by the second disturbance episode.

405

406 **3.3 The effect of past disturbance and future climate on 21st century carbon** 407 **sequestration**

408 Our simulations reveal a considerable impact of past disturbances on the current state of total
409 ecosystem carbon (Table 1). Simulations without disturbances resulted in an increase in carbon
410 storage of 43.9 tC ha⁻¹ (+11.0%) in 2013 compared to the baseline scenario (i.e., including
411 natural and human disturbance). The effect of disturbances was strongly dominated by forest
412 management (97.7%), with only a small influence of the two episodes of natural disturbance.
413 Past disturbances also resulted in a considerable carbon uptake beyond 2013 (Table 1, Fig. 5,
414 Fig. 6), *inter alia*, as a result of a persistent recovery of growing stock (Table 2). Past forest
415 management had a strong and continuous positive legacy effect on the future cumulative carbon
416 uptake of the landscape (cumulative decrease in NEE until 2099 of -40.6 tC ha⁻¹, p<0.001). The
417 second disturbance episode caused a release of carbon (positive NEE) over the first years of
418 future simulations, followed by a reversal of the trend towards a negative NEE effect (Fig. S14).
419 Its overall impact on cumulative NEE at the end of the simulation period was -3.5 tC ha⁻¹
420 (p=0.191), i.e. over the 21st century the recent disturbance period had an overall positive effect
421 on forest C sequestration. The first disturbance episode had almost no effect on the future
422 carbon dynamics (NEE effect of -0.2 tC ha⁻¹, p=0.792). Simulations of the total legacy effect
423 of past disturbances (both natural and human) resulted in a cumulative NEE of on average -43.8
424 tC ha⁻¹ (p<0.001) until 2099, indicating that a substantial future C uptake results from the
425 recovery of forest ecosystems from past disturbance (Fig. 6).



426 Climate change weakened the carbon sink strength on the landscape, mainly as a result of
427 climate-mediated differences in successional trajectories of forest ecosystems (Table 2).
428 However, climate change effects on NEE were more variable compared to disturbance legacy
429 effects, with increasing uncertainty over time as a result of differences in climate scenarios (Fig.
430 5). On average, climate change increased the cumulative NEE until 2099 by +24.0 tC ha⁻¹
431 ($p < 0.001$), and thus reduced the carbon uptake of the landscape relative to a continuation of
432 historic climate (Fig. 6).

433

434 **4. Discussion**

435 **4.1 Disturbance interactions in time**

436 Consistent with previous studies assessing the spatial and temporal autocorrelation of
437 disturbances in Europe (Marini et al., 2012; Schurman et al., 2018; Stadelmann et al., 2013;
438 Thom et al., 2013), we hypothesized that the disturbance episode in the early 20th century
439 influenced disturbances in the early 21st century. Our hypothesis was based on the importance
440 of landscape topography for wind and bark beetle disturbances (Senf and Seidl, 2018; Thom et
441 al., 2013), and the fact that susceptibility to these agents generally increases with stand age, and
442 is usually high after 90 years of stand development (Overbeck and Schmidt, 2012; Valinger and
443 Fridman, 2011). However, our analysis revealed a low probability for the same area to be
444 affected by the two consecutive disturbance episodes (Fig. 4). This finding is in contrast to
445 previous studies, which, however, investigated interactions between disturbance events in the
446 mountain forests of the Alps over only a few years (e.g., Pasztor and others 2014), while we
447 here analyzed temporal autocorrelation across multiple decades. Furthermore, also our focus on
448 an entire landscape (and its large heterogeneity in topographic settings and stand conditions) is
449 different from previous assessments of long-term disturbance feedbacks (but see Hanewinkel



450 et al., 2008), which have largely focused on plot to stand-level analyses using dendroecology
451 (e.g., Schurman et al., 2018).

452 We here tested for an amplifying feedback of natural disturbances in time, expecting high
453 susceptibility for large parts of the landscape recovering uniformly after the first disturbance
454 episode, and reaching high susceptibility to wind and bark beetles simultaneously. However,
455 disturbances can also have negative, dampening effects on future disturbance occurrence, e.g.,
456 when they lead to increased heterogeneity (Seidl et al., 2016) and trigger autonomous
457 adaptation of forests to new environmental conditions (Thom et al., 2017c). The low overlap
458 between the two disturbance episodes reported here could thus be an indication for such a
459 dampening feedback between disturbances, yet further tests are needed to substantiate this
460 hypothesis for Central European forest ecosystems. An alternative explanation for the diverging
461 spatial patterns of the two disturbance episodes might be a different wind direction in the storm
462 events initiating the two respective episodes, affecting different parts of the highly complex
463 mountain forest landscapes. Also the legacy effects from past forest management were different
464 for each episode. The more open structure within stands resulting from heavy exploitation
465 before 1900 may have increased wind susceptibility in the central and southern reaches of the
466 landscape regions.

467

468 **4.2 The role of disturbance legacies on future C uptake**

469 Past studies investigating drivers of the forest carbon balance have largely focused either on
470 historic factors (Keenan et al., 2014; Naudts et al., 2016) or future changes in the environment
471 (Manusch et al., 2014; Reichstein et al., 2013). Only few studies to date have explicitly
472 considered disturbance legacies when assessing climate change impacts on the future carbon
473 uptake of forest ecosystems. However, disregarding legacy effects could lead to a misattribution



474 of future forest C changes. Here we harnessed an extensive long-term documentation of
475 disturbance history to study impacts of past disturbance and future climate on the future NEE
476 of a forest landscape. We found long-lasting legacy effects of past disturbances on the forest
477 carbon cycle (see also Kashian et al., 2013; Landry et al., 2016; Nunery and Keeton, 2010),
478 supporting our hypothesis regarding the paramount importance of disturbance legacies for
479 future C dynamics. In line with a dynamic landscape simulation study for western North
480 America (Loudermilk et al., 2013) our results revealed that disturbance legacies have a stronger
481 effect on NEE than changes in climatic conditions (on average 1.7 times higher cumulative
482 effect over the 21st century – see Fig. 6). Disregarding legacy effects may thus cause a
483 substantial bias when studying the future carbon dynamics of forest ecosystems. It has to be
484 noted, however, that over longer future time frames as the one studied here the effects of climate
485 change will become more important relative to past legacy effects. While we here focused on
486 the strength of the disturbance legacy effect, future efforts could aim at determining its duration.
487 Moreover, while our analyses addressed the effects of wind and bark beetles – currently the two
488 most important natural disturbance agents in Central Europe (Thom et al., 2013) – as well as
489 their interactions, future climate change may increase the importance of other disturbance
490 agents not investigated here (see e.g., Wingfield et al., 2017).

491 The specific disturbance history of our study area, characterized by intensive natural and human
492 disturbances in the past and major socio-ecological transitions throughout the 20th century, is
493 key for interpreting our findings. In particular, the cessation of forest management in 1997 had
494 a very strong impact on the future carbon balance of the landscape (an on average 166.8 times
495 and 11.5 times higher effect than the first and second episodes of natural disturbances,
496 respectively – see Fig. 6). In addition to disturbance legacy effects, also climate change
497 significantly affected the future NEE. In contrast to the general notion that temperate forests
498 will serve as a strong carbon sink under climate change (Bonan, 2008), our dynamic simulations



499 suggest that climate change will decrease the ability of the landscape to sequester carbon in the
500 future, mainly by forcing a transition to forest types with a lower carbon storage potential.
501 However, considerable uncertainties of climate change impacts on the carbon balance of forest
502 ecosystems remain (e.g., Manusch et al., 2014). These uncertainties may arise from a wide
503 range of potential future climate trajectories, but also from a limited understanding of processes
504 such as the CO₂ fertilization effect on forest C uptake (Kroner and Way, 2016; Reyer et al.,
505 2014). In addition to the direct impacts of climate change (e.g., via temperature and
506 precipitation changes) on forest ecosystems, climate change will also alter future natural
507 disturbance regimes (Seidl et al., 2017). The potential for such large pulses of C release from
508 forests is making the role of forests in climate mitigation strategies highly uncertain (Kurz et
509 al., 2008; Seidl et al., 2014a).

510

511 **5. Conclusions**

512 Past disturbance (both human and natural) have a long-lasting influence on forest dynamics. In
513 order to project the future of forest ecosystems we thus need to better understand their past. We
514 here showed how a combination of historical sources and simulation modeling – applied by an
515 interdisciplinary team of scientists – can be used to improve our understanding of the long-term
516 trajectories of forest ecosystems (Bürgi et al., 2017; Collins et al., 2017; Deng and Li, 2016).
517 Two conclusions can be drawn from the strong historical determination of future forest
518 dynamics: First, as temperate forests have been managed intensively in many parts of the world
519 (Deng and Li, 2016; Foster et al., 1998; Naudts et al., 2016), their contribution to climate change
520 mitigation over the coming decades is likely determined already to a large degree by their past
521 (see also Schwaab et al., 2015). This means that for the time frame within which a
522 transformation of human society needs to be achieved in order to retain the earth system within



523 its planetary boundaries (Steffen et al., 2011), the potential for influencing the role of forests
524 might be lower than frequently assumed. Efforts to change forest management now to mitigate
525 climate change through *in situ* C storage, have high potential (Canadell and Raupach, 2008),
526 but will likely unfold their effects too late to make a major contribution to the transition of the
527 coming decades. Second, any changes in the disturbance regime of forests – whether intentional
528 (when altering management) or unintentional in the case of changing natural disturbances –
529 have profound consequences for the future development of forest ecosystems. This underlines
530 that a long-term perspective integrating past and future ecosystem dynamics is important when
531 studying forests, and that decadal to centennial foresight is needed in ecosystem management.

532

533 **Author contribution**

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

537

538 **Competing interests**

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

540

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547

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836

837 **Tables**

838 Table 1. Development of total ecosystem carbon stocks (tC ha^{-1}) over time and in different scenarios of disturbance history and future climate.
 839 Values are based on iLand simulations and indicate means and standard deviations (SD) over averaged landscape values for the respective
 840 scenarios. “Historic climate” assumes the continuation of the climate 1950 – 2010 throughout the 21st century, while “Climate change” denotes
 841 the effect of three alternative climate change scenarios for the 21st century. The first three columns indicate the respective permutation of the
 842 simulated disturbance history (see also Fig. 3), with the first line representing the historical reconstruction of landscape development. Y=yes,
 843 N=no.

First nat. dist. episode	Mgmt	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	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	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	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	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	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	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	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	333.0	0.1	458.6	0.5	471.7	0.6	487.9	0.9	464.1	21.0	

844

845



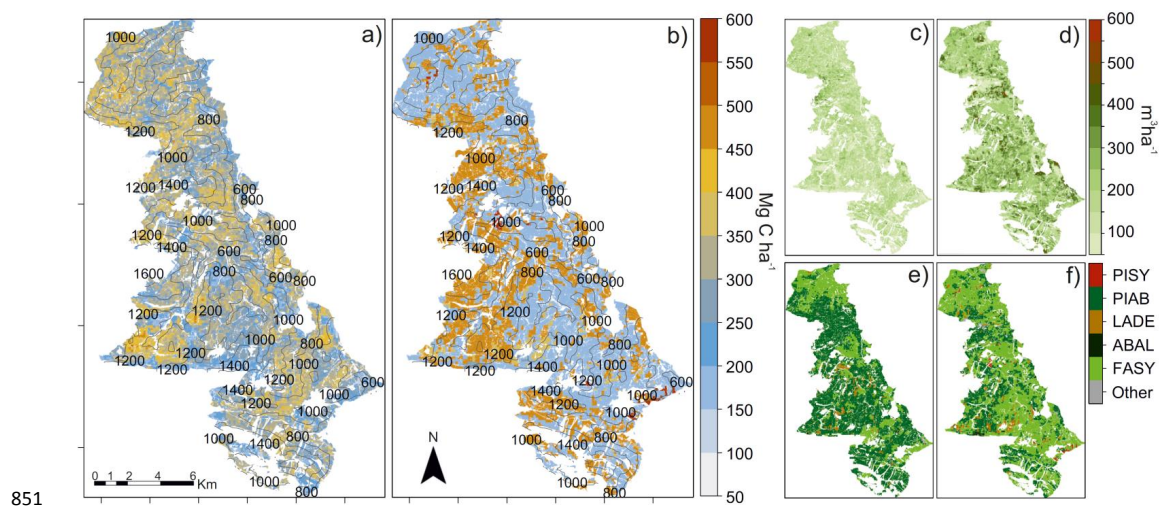
846 Table 2. Growing stock by tree species ($\text{m}^3 \text{ha}^{-1}$). Values are based on all iLand simulation runs and indicate species means and standard
 847 deviation (SD) over averaged landscape values. “Historic climate” assumes the continuation of the climate 1950 – 2010 throughout the 21st
 848 century, while “Climate change” denotes 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	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	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	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	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	6.0	0.2	14.7	1.4	16.0	1.6	13.4	0.5	23.8	1.7	
Total	212.3	247.4	0.8	467.4	79.0	518.5	66.0	622.6	35.4	613.3	46.5	

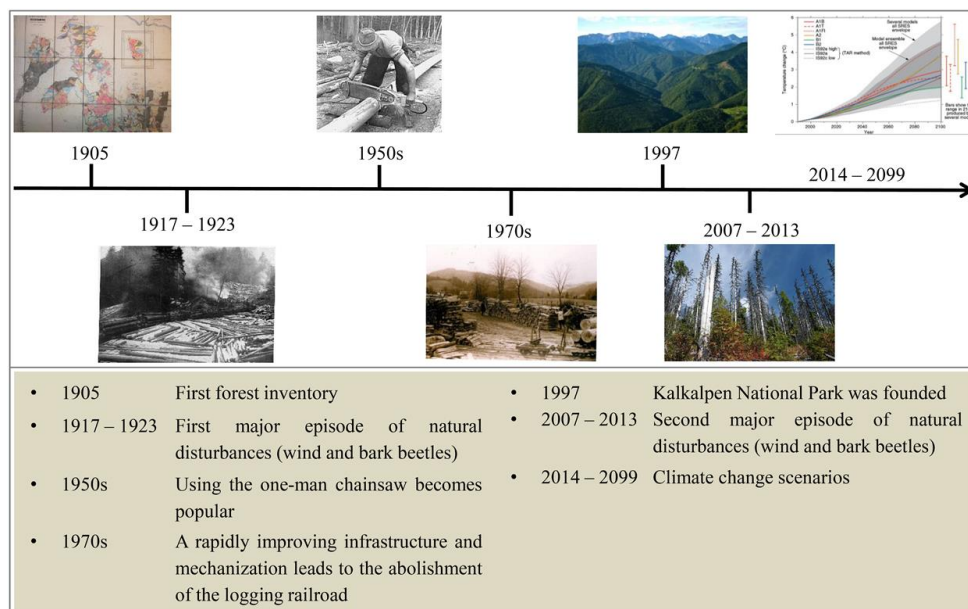
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850 **Figures**



852 Fig. 1: State of forest ecosystem attributes across the study landscape in 1905 and 2013. (a) and (b) show the distribution of total ecosystem
853 carbon, (c) and (d) present the growing stock, and (e) and (f) indicate the dominant tree species (i.e., the species with the highest growing stock)
854 in 1905 and 2013, respectively. PISY = *Pinus sylvestris*, PIAB = *Picea abies*, LADE = *Larix decidua*, ABAL = *Abies alba*, FASY = *Fagus*
855 *sylvatica*, and “Other” refers to either other dominant species, not individually listed here due to their scarcity, or areas where no trees are
856 present. Isolines represent elevational gradients.

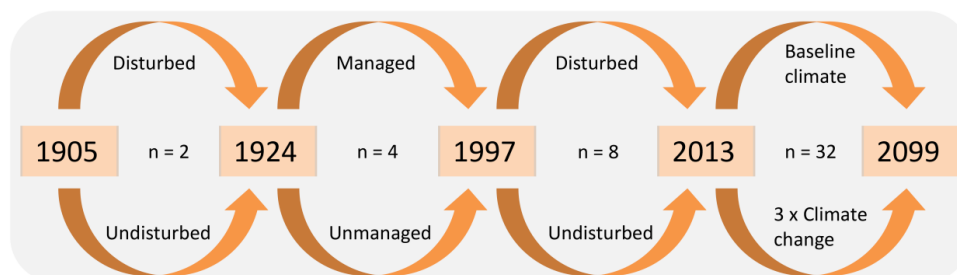


857

858 Fig. 2. Timeline of important historic events of relevance for the simulation of the study
 859 landscape. Timeline figures originate from various sources. 1905 and 1917 – 1923: archives of
 860 the Austrian Federal Forests; 1950s: <https://waldwissen.at>; 1970s: <https://atterwiki.at>; 1997:
 861 <http://kalkalpen.at>; 2007 – 2013: photo taken by the authors of this study; 2014 – 2099:
 862 <http://climate-scenarios.canada.cau>.



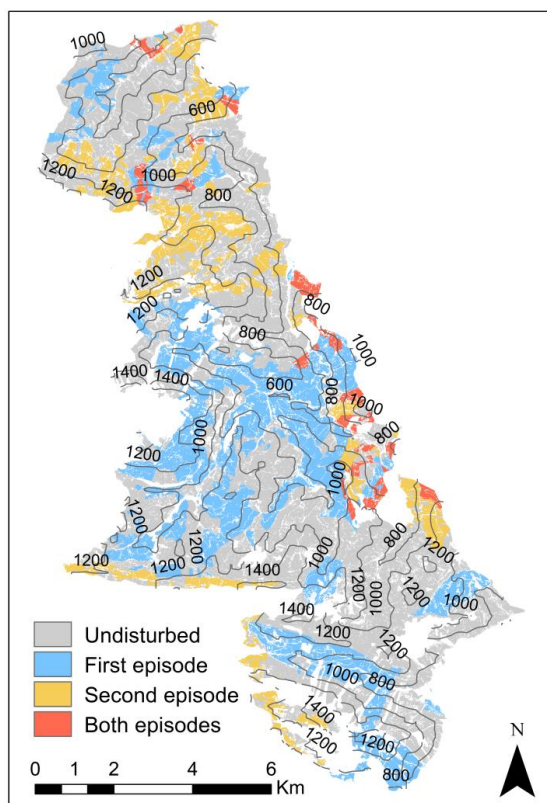
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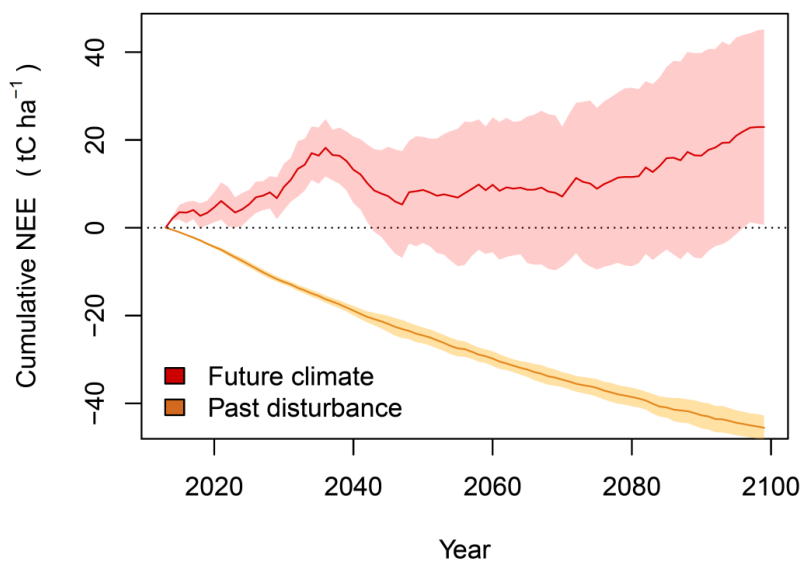
865 Fig. 3: The disturbance histories and climate futures considered in the simulation. The figure
866 shows the permutation of factors considered between each time step (years in boxes). n denotes
867 the number of unique combinations trajectories resulting from the addition of each individual
868 permutation, each of which was replicated 20 times.

869



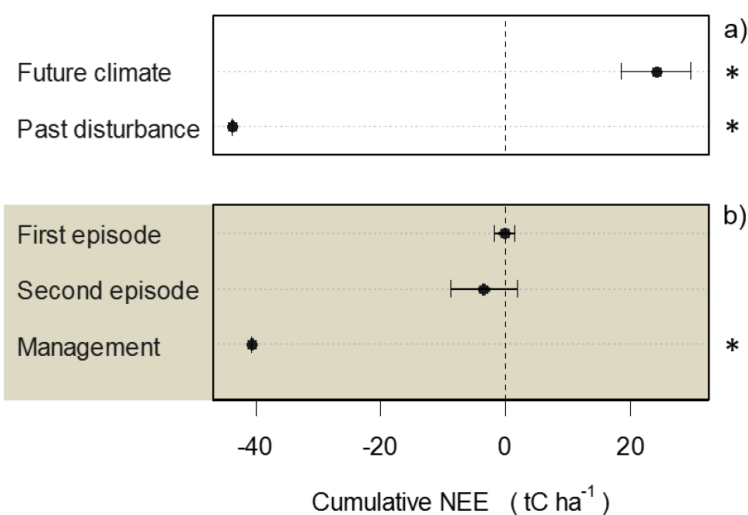
870

871 Fig. 4: Disturbance activity in two episodes of natural disturbance 1917 – 1923 (first episode)
872 and 2007 – 2013 (second episode). Isolines represent elevational gradients.



873

874 Fig. 5. Mean cumulative change in net ecosystem exchange (NEE) derived by comparing NEE
875 outputs including past disturbance (historic management and two episodes of natural
876 disturbance) and future climate with all scenarios excluding past disturbance and baseline
877 climate, respectively. Shaded areas denote the standard deviation in NEE for the respective
878 scenarios. NEE is the carbon flux from the ecosystem to the atmosphere.



879

880 Fig. 6. Effects of future climate and past disturbance on the cumulative NEE of the period 2014
 881 – 2099. a) Effect sizes are calculated from a comparison between climate change and historic
 882 climate (both without disturbance) as well as disturbed and undisturbed scenarios (both under
 883 historic climate conditions), respectively. Whiskers give the 95% confidence interval around
 884 the effect size, and asterisks indicate significant indicators ($\alpha=0.05$). b) In addition to the overall
 885 effect of past disturbance, the effect was subdivided into the first and second episodes of natural
 886 disturbance as well as human-induced disturbance via management (shaded box).