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Effects of using certain tree species in forest regeneration on regional wind damage risks in Finnish boreal forests under different CMIP5 projections

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Abstract

We studied how the use of certain tree species in forest regeneration affected the regional wind damage risks to Finnish boreal forests under the current climate (1981–2010) and recent-generation global climate model (GCM) predictions (i.e., 10 GCMs of CMIP5, with wide variations in temperature and precipitation), using the representative concentration pathways RCP4.5 and RCP8.5 over the period 2010-2099. The study employed forest ecosystem and mechanistic wind damage risk model simulations on upland national forest inventory plots throughout Finland. The amount of wind damage was estimated based on the predicted critical wind speeds for uprooting trees and their probabilities. In a baseline management regime, forest regeneration was performed by planting the same tree species that was dominant before the final cut. In other management regimes, either Scots pine, Norway spruce or silver birch was planted on medium-fertility sites. Other management actions were performed as for a baseline management. The calculated amount of wind damage was greatest in southern and central Finland under CNRM-CM5 RCP8.5, and the smallest under HadGEM2-ES RCP8.5. The most severe climate projections (HadGEM2-ES RCP8.5 and GFDL-CM3 RCP8.5) affected the wind damage risk even more than did the tree species preferences in forest regeneration. The situation was the opposite for the less severe climate projections (e.g., MPI-ESM-MR RCP4.5 and MPI-ESM-MR RCP8.5). The calculated amount of wind damage was clearly greater in the south than in the north, due to differences in forest structure. The volume of growing stock is much higher in the south for the more vulnerable Norway spruce (and birch) than in the north, which is opposite for the less vulnerable Scots pine. The increasing risk of wind damage should be taken into account in forest management because it could amplify, or even cancel out, any expected increases in forest productivity due to climate change.

Keywords Climate change \cdot Forest management \cdot Gap-type forest ecosystem model \cdot Mechanistic wind damage model \cdot RCP4.5 \cdot RCP8.5 \cdot Tree species preference \cdot Wind damage

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Introduction

Since the 1990s, strong winds and storms have caused large economic losses to forestry in central and northern Europe (Schelhaas et al. 2003; Gardiner et al. 2010; Schuck and Schelhaas 2013). In northern Europe, and in forested countries like Finland, most wind damage has occurred in stands adjacent to newly clear-cut areas, or in recently heavily thinned older stands (Laiho 1987; Zubizarreta-Gerendiain et al. 2012; Suvanto et al. 2016). During the coming decades, the risk of wind damage to forests is expected to increase, although the frequency and severity of the storms may not increase (Nikulin et al. 2011; Pryor et al. 2012; Outten and Esau 2013; Mölter et al. 2016). This is due to the reduced period of frozen soil and tree anchorage in winter (Peltola



et al. 1999a; Kellomäki et al. 2010; Gregow et al. 2011a; Lehtonen et al. 2018). The soil is expected to barely freeze at all, for example, in southern and central Finland by 2100 under severe climate warming.

In Finland, about 45% of the volume of growing stock is currently accounted for by Scots pine (*Pinus sylvestris* L.), 31% by Norway spruce [*Picea abies* (L.) Karst.] and 24% by silver and downy birch (*Betula pendula* Roth and *Betula pubescens* Ehrh.) and other broadleaves (Finnish Forest Research Institute 2014). The increased cultivation of Norway spruce and its proportion of the growing stock volume may greatly increase the risk of wind damage to forests in Finland because Norway spruce, with its shallow rooting system, is more vulnerable to uprooting (i.e., lower critical wind speeds would be needed) than Scots pine and birch (Peltola et al. 1999b, 2010). Also, in central Europe, Norway spruce has already suffered the most wind damage among the conifers (Schmidt et al. 2010; Reyer et al. 2017).

Climate change is expected to increase the productivity of forests, especially in the northern boreal zone, due to improving growing conditions (Bergh et al. 2003; Briceño-Elizondo et al. 2006; Koca et al. 2006; Kellomäki et al. 2008, 2018; Poudel et al. 2011). At the same time, it may decrease the productivity in the southern boreal zone (Koca et al. 2006; Kellomäki et al. 2008, 2018; Reyer et al. 2017). The responses of forests to climate change may differ greatly at the regional level. This is related to differences in the prevailing environmental conditions (climate, site), current forest structure (age, species) and forest management regimes (Bergh et al. 2003; Briceño-Elizondo et al. 2006; Garcia-Gonzalo et al. 2007; Kellomäki et al. 2008; Lindner et al. 2010; Alrahahleh et al. 2018). Particularly under severe climate warming, the growth and success of Norway spruce are expected to decrease in Finland, especially on southern upland forest sites, if the growing conditions (temperature and water availability) become suboptimal for its growth (Briceño-Elizondo et al. 2006; Jyske et al. 2010; Kellomäki et al. 2018; Ruosteenoja et al. 2018). Increasing forest disturbances, such as wind damage, may also amplify, or even cancel out, any expected increases in the productivity of forests under changing climate (Kellomäki et al. 2008; Reyer et al. 2017).

According to Ikonen et al. (2017), the increased cultivation of Norway spruce under changing climate will increase the wind damage risk and the amount of damage in Finland, in the long run. Furthermore, tree species preference in forest regeneration may affect the wind damage risk more than the climate change, based on recent-generation global climate model (GCM) predictions (i.e., multi-model means of the Coupled Model Intercomparison Project Phase 5—CMIP5),

using the representative concentration pathways RCP4.5 and RCP8.5 over the period 2010–2099 (Ruosteenoja et al. 2016). Based on these multi-model means for RCP4.5 and RCP8.5, mean temperature and precipitation are expected to increase in Finland, depending on geographical region, by an average of 3–5 °C and 7–14%, respectively, during April–September by 2070–2099. Concurrently, atmospheric $\rm CO_2$ concentrations are expected to increase from the current value of 360 to 536 ppm (RCP4.5) and 807 ppm (RCP8.5) during the period 2070–2099.

Apart from the multi-model means for RCP4.5 and RCP8.5, certain individual GCMs, such as GFDL-CM3 RCP8.5 and HadGEM2-ES RCP8.5, have predicted up to a 6–7 °C increase in temperature during the potential growing season by 2070–2099, depending on geographical region in Finland. At the same time, they predict a slight to moderate increase in precipitation in the north, but only a slight increase (GFDL-CM3 RCP8.5), or a decrease (HadGEM2-ES RCP8.5), in precipitation in the south. Under such severe climate change outlooks, an increased cultivation of Norway spruce may decrease the growth and volume of growing stock, especially under southern boreal conditions, more than predicted, based on the multi-model means of the GCMs (Alrahahleh et al. 2018). Consequently, the severe climate projections (HadGEM2-ES RCP8.5 and GFDL-CM3 RCP8.5) may affect the wind damage risk even more than the tree species preferences in forest regeneration, which should be considered in forest management decision making.

Depending on the severity of the climate change projection, radically different adaptive measures for forestry might be useful. For example, Norway spruce could grow well on certain sites under less severe climate projections and thus would be preferred in forest regeneration, but this would not be the case for more severe climate projections (Alrahahleh et al. 2018). Therefore, the use of different individual GCM projections under RCPs is required in considering the uncertainties in the model predictions for forest growth and dynamics and consequently for the wind damage risks to forests. This is needed in order to properly adapt forest management to climate change. The forest ecosystem models, together with up-to-date information on current forest resources, and different climate projections offer a means of predicting forest growth and dynamics under changing management regimes and environmental conditions, respectively (Garcia-Gonzalo et al. 2007; Seidl and Lexer 2013; Alrahahleh et al. 2017, 2018; Reyer et al. 2017). The use of the simulated outputs of forest ecosystem models as inputs for mechanistic wind damage models also offers a means of predicting the threshold wind speeds needed for wind damage to forests and, consequently, their probabilities and the



amount of damage to expect (Gardiner et al. 2008; Peltola et al. 2010; Seidl et al. 2014; Ikonen et al. 2017).

In this context, we studied how the use of certain tree species in forest regeneration could affect the regional wind damage risks to Finnish boreal forests under the current climate (1981–2010) and recent-generation global climate model (GCM) predictions (i.e., 10 GCMs of CMIP5, with wide variations in temperature and precipitation; Ruosteenoja et al. 2016), using RCP4.5 and RCP8.5 over the period 2010-2099. The study employed forest ecosystem (SIMA; Kellomäki et al. 2005, 2008, 2018) and mechanistic wind damage (HWIND; Peltola et al. 1999b) model simulations on upland National Forest Inventory (NFI) plots throughout Finland. In a baseline management regime, forest regeneration was performed by planting the same tree species that was dominant before the final cut. In other management regimes, either Scots pine, Norway spruce or silver birch was planted on medium-fertility sites. Other management actions concerning rotation were performed as for baseline management.

We hypothesized that the use of certain individual GCM runs of CMIP5 could lead to contradictory results for the expected wind damage risks to those predicted based on the multi-model means under RCP4.5 and RCP8.5 by Ikonen et al. (2017). In addition, we expected that very severe climate change projections could affect the wind damage risk to a greater degree than tree species preferences in forest regeneration. The same simulation layout was recently used by Alrahahleh et al. (2018), who studied how the use of certain tree species in forest regeneration affected the volume growth, timber yield and carbon stock of boreal forests in Finland under different CMIP5 projections.

Material and methods

Initial stand and site data

The initial stand and site characteristics used in the simulations were based on the 10th National Forest Inventory of Finland (see Korhonen 2016 for more details). One randomly selected sample plot from every permanent cluster of sample plots on upland forest land assigned to timber production throughout Finland was used (see Table 1 for more details). Altogether, the data included 2642 sample plots, of which most—1388 plots—were on medium-fertile mesic sites (*Myrtillus* type, MT), 529 were on fertile herbrich (*Oxalis myrtillus* type) or more fertile sites, 641 were on less fertile subxeric sites (*Vaccinium* type), and 84 were on poorer, dryish sites (*Cladonia* type).

Climate data

The climate data for the current climate were based on measurements made by the Finnish Meteorological Institute (FMI) of temperature and precipitation over the period 1981–2010. For the climate change projections, we used the results from 10 individual GCMs (four GCMs driven by both RCP4.5 and RCP8.5 and two additional GCMs driven by RCP8.5 only), which were downloaded from the CMIP5 database by the FMI (Fig. 1, Table 2; Ruosteenoja et al. 2016). These individual GCMs give very different climate projections, even under the same radiative forcing scenario (e.g., RCP8.5). The reasons for this may be many, such as the GCMs being produced by different research institutes (and countries of origin) differing in terms of model parameterization and structure, using different input datasets, spatial resolution and numerical algorithms (see Ruosteenoja et al. 2016). For comparison, in this study, we used the multimodel mean values of 28 individual GCMs under RCP4.5 and RCP8.5, which are the same as those used in a study by Ikonen et al. (2017).

The 10 selected GCMs, which have a proven ability to relatively accurately simulate the temperature and precipitation of the current climate (1981–2010) in northern Europe (Lehtonen et al. 2016a, 2016b; Ruosteenoja et al. 2016), provided us with a good representation of the overall variability in the full ensemble of CMIP5 projections for monthly mean temperatures and precipitation for 2010–2099. Too high or low predicted values for daily mean temperatures and precipitation, in relation to the observed data, however, still needed to be bias-corrected, which was done using an empirical bias correction method called quantile mapping (see for more details, Räisänen and Räty 2013; Räty et al. 2014). Both the observational and climate change data were interpolated by the FMI onto a 10×10 km grid throughout Finland, using the kriging with external drift method (see for more details, Venäläinen et al. 2005; Aalto et al. 2013, 2016), before they were used in the simulations.

Management activities

In our forest ecosystem model simulations under a baseline management regime, forest regeneration was always done by planting the same tree species that was dominant before the final cut. In other management regimes, either Scots pine, Norway spruce or silver birch were planted on mediumfertility (MT) sites, as all these tree species are suitable for such sites. On other site types, baseline management was followed. Other management actions concerning rotation were performed as for the baseline management (Table 1).



Table 1 Simulation layout with stand and site conditions, climate projections and management activities (same simulation layout was used also in the study of Alrahahleh et al. 2018)

Simulation layout	Description
Initial stand and site conditions	The initial stand and site characteristics represented one randomly selected sample plot of NFI10 from every permanent cluster of sample plots on upland forest land assigned to timber production. Southern Finland denoted for the area of old administrational Forest Centre units 1–6, central Finland units 7–10 and northern Finland units 11–13, respectively. The average distance between the clusters of sample plots is 6×6 km in units 1–12 and 10×10 km in unit 13. Each sample plot had on average of nine trees, for which tree species and diameter at breast height (dbh, cm) were available
Climate data	Current climate data (1981–2010), four GCMs driven by both RCP4.5 and RCP8.5 and two additional GCMs driven by RCP8.5 only, and multi-model mean values driven by both RCP4.5 and RCP8.5 forcing scenarios (2010–2099)
Species-specific response to the temperature sum	Minimum (TS_{min} , 370, 390, 390 degree days, d.d.), maximum (TS_{max} , 2060, 2500, 4330 d.d.) and optimum (TS_{opt} , 1215, 1445, 2360 d.d.) temperature sum values for growth were smallest in Norway spruce, followed by Scots pine and birch
Soil moisture availability	The field capacity and wilting point defined the soil moisture available for tree growth on different site and soil types as affected by precipitation and evaporation. Scots pine was more drought tolerant than other tree species
Initial amount of soil organic matter (and carbon) and nitrogen	The initial amounts of soil organic matter (and carbon) and nitrogen available for growth were defined based on the site fertility type and regional temperature sum of the current climate. Atmospheric nitrogen deposition of 10 kg year ⁻¹ was assumed (Järvinen and Vänni1994; Kellomäki et al. 2005)
Forest regeneration	In a baseline management, it was planted the same tree species that was dominant before the final cut. In alternative regimes, either Scots pine, Norway spruce or silver birch was planted on medium-fertile (MT) sites. In planting it was used 2000 seedlings ha ⁻¹ for Norway spruce and Scots pine, and 1600 seedlings ha ⁻¹ for silver birch (diameter of 2.5 cm). In addition, seedlings were expected to regenerate naturally at all sites
Tending of seedling stand	Tending of the seedling stand was done before the first commercial thinning by removing mostly smaller or suppressed trees
Thinnings and final felling	The region-, site- and tree species-specific management recommendations were used as a basis for the timing and intensity of thinnings (from below), and timing of final felling. When the basal area threshold for thinning was reached, at a given dominant height, the thinning could be done by reducing the basal area to the recommended threshold value after thinning. Final felling was performed based on the basal area weighted diameter at breast height (with a range 22–30 cm depending on region, site and tree species). However, on average, a 13-year delay in harvesting was used, compared to the management recommendations
Harvesting intensity	Only timber (sawlogs and pulpwood with minimum top diameters of 15 cm and 6 cm) was harvested, and logging residues were left at the site
Other information	Parts of plots from central (10%) to northern (30%) Finland were left outside management, unlike in southern Finland, where the current forest conservation area is very small, at around 2% (Finnish Forest Research Institute 2014)

As a basis for the tending of seedling stands, thinnings and final fellings, region-, site- and tree species-specific management rules were used (Äijälä et al. 2014). On the other hand, on average, a 13-year delay in thinnings and final fellings was used, compared to the management recommendations, because these are often delayed in practice (see Finnish Forest Research Institute 2014). Parts of the forest plots from central to northern Finland were also left outside of management, unlike in southern Finland, where currently the total forest conservation area is very small (Finnish Forest Research Institute 2014). This has been done previously by

Ikonen et al. (2017) and Alrahahleh et al. (2018), resulting in more realistic predictions for the growth and volume of growing stock for the first 30-year simulation period (2010–2039) under current climate, compared to the forest statistics for the period 2004–2009 (Finnish Forest Research Institute 2014).

Outlines for the forest ecosystem model

A gap-type forest ecosystem model SIMA (Kellomäki et al. 2005, 2008, 2018) was used to simulate the regeneration,



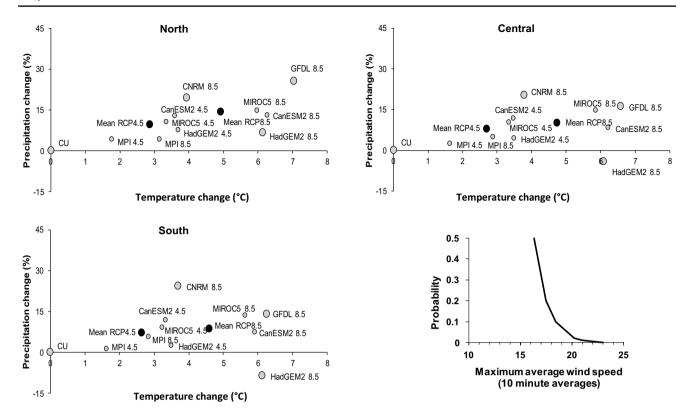


Fig. 1 Climate change projections with temperature and precipitation change in April–September in the third period, 2070–2099, in northern, central and southern Finland (average of the 30-year period). Black circles represent the climate change projections 'Mean RCP4.5' and 'Mean RCP8.5' (means of 28 individual model runs),

and gray circles represent individual climate change projections and the current climate. The changes are relative to the baseline climate (1981–2010). Also, the probabilities of 10 min (measured) maximum average wind speeds in Helsinki-Vantaa Airport weather station are presented (based on Peltola et al. 2010)

Table 2 Mean changes in temperature (ΔT , °C) and precipitation (ΔP , %) under different CMIP5 projections (i.e., multi-model means and individual GCMs) during potential growing seasons (April–September) in the period 2070–2099 in southern (old administrational

Forest Centre Units 1–6) and northern (11–13) Finland, in comparison to current climate (1981–2010, with a mean atmospheric CO_2 concentration of 360 ppm)

Climate model acronym	Institution/country of origin	ΔT (°C))	$\Delta P\left(\%\right)$	
(short name in bold)		South	North	South	North
HadGEM2-ES RCP4.5	Met Office Hadley Centre for Climate Science and Services, UK	3.5	3.7	2	8
HadGEM2-ES RCP8.5		6.1	6.1	-9	7
MPI-ESM-MR RCP4.5	Max Planck Institute for Meteorology, Germany	1.6	1.8	1	4
MPI-ESM-MR RCP8.5		2.8	3.1	6	4
CanESM2 RCP4.5	Canadian Centre for Climate Modelling and Analysis, Canada	3.3	3.6	12	13
CanESM2 RCP8.5		5.9	6.3	7	13
MIROC5 RCP4.5	Atmosphere and Ocean Research Institute (University of Tokyo),	3.2	3.3	9	11
MIROC5 RCP8.5	National Institute for Environmental Studies and Japan Agency for Marine-Earth Science and Technology, Japan	5.6	6	13	15
CNRM-CM5 RCP8.5	National Center for Meteorological Research, France	3.7	3.9	24	19
GFDL-CM3 RCP8.5	NOAA Geophysical Fluid Dynamics Laboratory, USA	6.3	7	14	26
Mean (28) GCMs, RCP4.5	_	2.6	2.9	7	10
Mean (28) GCMs, RCP8.5	-	4.6	4.9	9	14

The predicted mean atmospheric CO₂ concentrations under the RCP4.5 and RCP8.5 forcing scenarios were 536 ppm and 807 ppm, respectively, for the period 2070–2099. Other information for individual GCMs available in Ruosteenoja et al. (2016)



growth and mortality of trees in boreal upland forests (on mineral soils) throughout Finland. Under optimal conditions, the growth and/or regeneration are not assumed to be limited by temperature sum (TS > 5 °C threshold), light availability, soil moisture or nitrogen supply. In addition, growth is affected by atmospheric $\rm CO_2$ concentration, atmospheric nitrogen deposition and tree maturity (diameter at 1.3 m above the ground). The tree diameter is further used to calculate tree height and the mass of tree organs (foliage, branches, stem and roots).

The species-specific response to the temperature sum was modeled for the main Finnish boreal tree species, based on a downward-opening symmetrical parabola (Kienast 1987; Nikolov and Helmisaari 1992; Kellomäki et al. 2008, 2018). This was done by assuming that the minimum and maximum values of the temperature sum define the geographical distribution of each tree species throughout the boreal zone (Table 1). In calculating the effects of changing climate on forest growth, only the changes in monthly temperature sums from April to September (i.e., the potential growing season), compared with the temperature sum of the current climate, were taken account (Torssonen et al 2015; Kellomäki et al. 2018). This was done in order to consider the prevailing light conditions.

Field capacity and wilting point define the soil moisture available for tree growth at different sites and soil types, as affected by precipitation and evaporation. Under optimum conditions, soil moisture is greater than the wilting point (no dry days). The initial amount of soil organic matter (and carbon) and nitrogen available for growth are defined based on the site fertility type and the regional temperature sum of the current climate (see Kellomäki et al. 2005, 2008). The amount of soil organic matter (and carbon) and nitrogen available for growth are also affected by the input of litter and deadwood to the soil layer, and their decay.

In our simulations, management included control over artificial regeneration (planting) with the desired spacing and tree species (including naturally born seedlings), over stand density in tending the seedling stand, and over thinning and the final cut (only timber was harvested in this study; Table 1). The initial properties of a tree stand were described in terms of tree species, including the number of trees per hectare in each diameter class. The model simulations were carried out with a time step of one year on an area of 100 m², based on the Monte Carlo technique (see, e.g., Bugmann et al. 1996), based on which certain

events, such as the birth and death of trees, are stochastic events. Therefore, only the mean tendency of several iterations was considered in further data analyses (e.g., 20 iterations in this study, for which a coefficient of variation for the volume of growing stock was 1.6% over a 90-year simulation period at the plot level, based on our calculations).

Previous results from model simulations have shown good agreement (a correlation of 0.857) with the measured average annual volume growth (1996–2003) of the main Finnish boreal tree species on the permanent upland National Forest Inventory plots for different regions of Finland (Kellomäki et al. 2008). Routa et al. (2011) also showed that the simulated mean annual volume growth of Norway spruce and Scots pine stands, on medium-fertility sites in 13 different locations throughout Finland, using the SIMA model and a statistical growth and yield model (MOTTI; Hynynen et al. 2002) indicated a good agreement between the SIMA and MOTTI simulations ($R^2 = 0.85$).

Outlines for a mechanistic wind damage model

In this work, the outputs of the SIMA model for different sample plots (i.e., tree species, tree height and diameter at breast height (DBH) for each sample tree and stand density) were used as inputs for a mechanistic wind damage model (HWIND; Peltola et al. 1999b), which predicts the critical wind speeds (CWSs, m s⁻¹) needed to uproot Scots pine, Norway spruce and birch trees in various stand configurations. The CWSs are computed at a height of 10 m above an open lawn surface (10 min averages; see Dupont et al. 2015). A tree is uprooted if its maximum bending moment exceeds the resistance of the root-soil plate, and breaks if it exceeds the resistance of the stem (Peltola et al. 1999b). In calculating the CWS for individual trees in a stand, the stand density and dominant stand height are used to calculate the mean wind profile, which is later applied to individual trees.

Based on previous HWIND simulations (see, e.g., Peltola et al. 1999b; Dupont et al. 2015), Norway spruce, with the shallowest rooting, has the lowest CWS, followed by birch (in leaf, in summer), and Scots pine (with the deepest rooting) with the highest CWS, using the same tree and stand characteristics. In autumn, without leaves, birch is supposed to have a very low/no risk because of its low surface area (i.e., very high CWS needed). The outputs of the HWIND model (i.e., the CWSs needed to uproot or break trees at the stand edge) have been in reasonable



agreement with other mechanistic wind damage models, such as GALES and FOREOLE (see Gardiner et al. 2000; Ancelin et al. 2004).

The properties of the HWIND model, its parameters, inputs and the validity of its outputs, as well as its performance for upland forests in Finland and Sweden, have been discussed in detail by, for example, Peltola et al. (1999b), Blennow and Sallnäs (2004), Zeng et al. (2006) and Gardiner et al. (2008). HWIND has also identified reasonably well the observed wind-damaged areas in previous case studies in Finland and Sweden (Talkkari et al. 2000; Blennow and Sallnäs 2004; Zubizarreta-Gerendiain et al. 2012).

In this study, the CWS calculations were performed by assuming unfrozen soil conditions and only considering the uprooting of trees. This was done because forests are vulnerable to wind damage in Finland mainly in unfrozen soil, and uprooting is the most common wind damage type. Under frozen soil conditions, stem breakage, which was not considered here, is the typical failure type (Peltola et al. 2000). It was also assumed that all trees at risk were located within one dominant stand height distance from the new upwind stand edge, where they have been observed to have the greatest risk of damage under Finnish conditions (Peltola et al. 1999b; Zubizarreta-Gerendiain et al. 2012). Thus, our CWS values represent the maximum vulnerability of trees to uprooting.

Critical wind speeds, their probabilities and amount of wind damage

Based on the calculated CWSs and their probabilities, we estimated the amount of wind damage using the approach outlined by Ikonen et al. (2017). First, we calculated the average minimum CWS for each sample plot (over all tree species in one simulation run), over each 30-year period (2010–2039, 2040–2069 and 2070–2099). The average minimum CWSs were also calculated separately, assuming either birch in leaf (summer) or leafless (autumn). Leafless birch has a very low surface area for wind to affect, so has a very low risk of wind damage. Trees with a height of < 10 m were also considered to have very low/no risk. Under Finnish conditions, forest stands are not liable to wind damage before the first commercial thinning (see, e.g., Zubizarreta-Gerendiain et al. 2012), which is done at a dominant height of about 12–16 m (see Äijälä et al. 2014).

The annual probabilities (dimensionless, in a range of 0–1) of the average minimum CWS were calculated throughout Finland using the probabilities of 10-min maximum average wind speeds, estimated by weather station at Helsinki Airport, i.e.,

$$P_{\text{cws}} = e^{y} / (1 + e^{y}) \tag{1}$$

where

$$y = 21.79 - 1.058 \times CWS \tag{2}$$

Peltola et al. (2010) and Zubizarreta-Gerendiain et al. (2017) (Fig. 1). This was done because the probabilities of strong wind speeds have been observed to be, on average, highest in southernmost Finland, followed by northern Finland (Peltola et al. 2010). Helsinki Airport data could also be considered to be highly representative for typical wind damage conditions in Finland, such as at the immediate downwind edges of new forest clear-cuts. Also, based on a preliminary analysis using different GCMs under different RCPs (FMI, unpublished results), this situation may not change in Finland, and wind conditions may not differ greatly under unfrozen or frozen soil conditions.

Based on the probabilities for the CWSs and the average volume of growing stock, we then estimated the volume of growing stock at risk (m³ ha⁻¹), and the amount of damage (m³ ha⁻¹) and percentage of damage to the volume of growing stock. In calculating the volume of growing stock at risk, and the amount of damage, we assumed that all sample trees were located within one stand height distance from the new vulnerable upwind edge, and that only a small proportion (3%) of the total stem volume would be damaged, based on the study by Zubizarreta-Gerendiain et al. (2012).

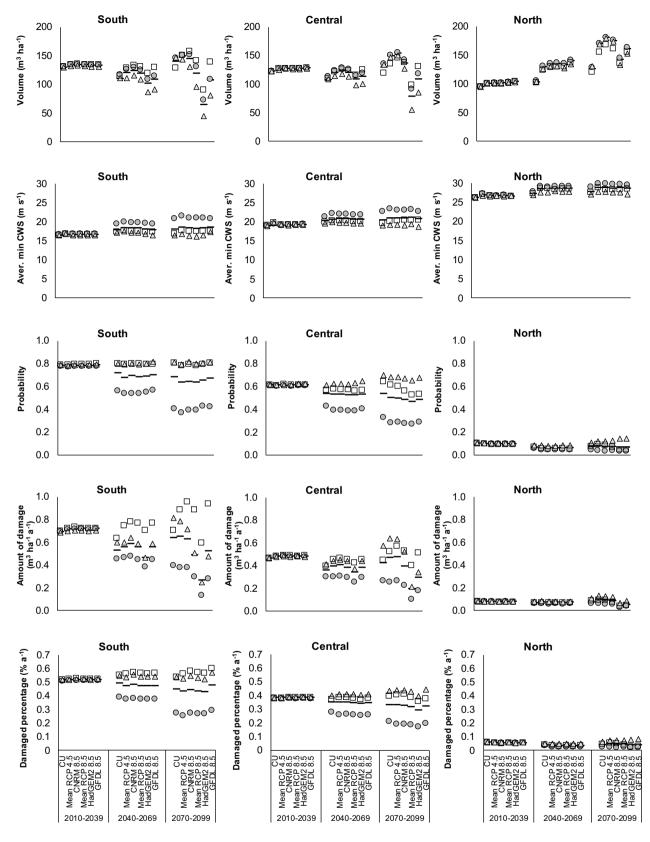
The results calculated at the plot level were then averaged for southern, central and northern Finland, over each 30-year period, for the current and all changing climate projections, under baseline management and increased use of different tree species in forest regeneration. The ranges of the results are shown for the different GCMs in order to demonstrate the sensitivity of the predictions to them. In addition, the results for current climate, the multi-model mean projections (mean RCP 4.5 and mean RCP 8.5), and the most extreme GCM projections (CNRM 8.5, HadGEM2 8.5 and GFDL 8.5) are discussed in more details below.

Results

Calculated average minimum CWSs and their probabilities

In the first 30-year period under the current climate, with baseline management, the predicted average minimum CWSs were smaller in summer (birch in leaf). They were also smaller in southern Finland (16 m s⁻¹) than in central







<Fig. 2 Average volume of growing stock (m³ ha⁻¹), average minimum CWS (m s⁻¹), predicted probabilities (dimensionless, in a range of 0–1) based on average minimum CWSs, amount of damage (m³ ha⁻¹ a⁻¹) and damaged percentage (% a⁻¹) in summer for each period and each management scenario under several climate change projections in southern, central and northern Finland. Medium-fertility sites were planted with Scots pine (Pref. SP), Norway spruce (Pref. NS) or silver birch (Pref. B), or with the tree species that was dominant before the final clear-felling (baseline management)

(19 m s⁻¹) and northern (27 m s⁻¹) Finland (Figs. 2, 3, Appendix Table 3). Also, under the 10 GCM projections in the third 30-year period, they were smallest in southern Finland. They were the smallest with an increased use of Norway spruce ($16-17 \text{ m s}^{-1}$), followed by an increased use of birch ($17-18 \text{ m s}^{-1}$), baseline management ($18-19 \text{ m s}^{-1}$), and an increased use of Scots pine ($20-21 \text{ m s}^{-1}$). In central and northern Finland, the tendency of the CWS ranges was similar to that in southern Finland, but with ranges of $18-23 \text{ m s}^{-1}$ and $27-30 \text{ m s}^{-1}$, respectively.

Under the current climate, with baseline management, the probability of wind damage in the summer was the highest in southern (0.8), followed by central (0.6) and northern (0.1) Finland, in the first 30-year period (Fig. 2, Appendix Table 4). In the third 30-year period, the probability of wind damage was in the range of 0.4–0.8, 0.3–0.7 and 0.0–0.1 in southern, central and northern Finland, respectively, under the 10 GCM projections. The probabilities were the lowest with an increased use of Scots pine, and the highest with an increased use of both Norway spruce and birch, and having marginal differences between each GCM projection. Generally, CWSs were higher in autumn (leafless birch) than in summer, regardless of management regime or climate projection (Figs. 3, 4).

Calculated amount of wind damage

Under the current climate, with baseline management, the calculated average amount of wind damage in summer was highest in southern (0.7 m³ ha⁻¹ a⁻¹), followed by central (0.5 m³ ha⁻¹ a⁻¹) and northern (0.1 m³ ha⁻¹ a⁻¹) Finland, in the first 30-year period (Figs. 2, 5, Appendix Table 5). In the third 30-year period, under 10 GCM projections, the calculated average amount of wind damage was in the range of 0.3–0.7 m³ ha⁻¹ a⁻¹ in southern Finland, when baseline management scenario was applied. With an increased use of Scots pine, the corresponding range was 0.1–0.4 m³ ha⁻¹ a⁻¹ (the smallest), whereas with an increased use of Norway spruce, it was 0.2–0.8 m³ ha⁻¹ a⁻¹, and with an increased use of birch, it was 0.6–1.0 m³ ha⁻¹ a⁻¹ (the largest). In central Finland,

the calculated amount of damage was in the range of $0.1-0.7~{\rm m}^3~{\rm ha}^{-1}~{\rm a}^{-1}$. There, in general, it was the smallest with an increased use of Scots pine and the greatest with an increased use of Norway spruce and birch. In northern Finland, the calculated amount of damage was marginal.

In general, both the increased use of a certain tree species and the intensity of climate change affected the amount of damage. In southern and central Finland, the calculated amount of damage was the greatest under CNRM 8.5 (similar to mean RCP 4.5) and the smallest under HadGEM2 8.5; however, the most extreme climate change projections (HadGEM2 8.5 and GFDL 8.5) affected the amount of wind damage risk even more than did tree species preference in forest regeneration, opposite to the mild climate change scenario (such as MPI 4.5 and MPI 8.5).

Under the current climate, with baseline management, the damage percentage (relative share of damage to the average volume of growing stock) in summer was greatest in southern (0.5%), followed by central (0.4%) and northern (0.1%) Finland, in the first 30-year period (Fig. 2). In the third 30-year period, under 10 GCM projections, the corresponding range was 0.3–0.6% in southern and 0.2–0.4% in central Finland, whereas it was marginal in northern Finland. The damage percentages were the lowest with an increased use of Scots pine, and the highest with an increased use of Norway spruce and birch. In general, individual GCM projections affected the damage percentages much less than the preference for different tree species in forest regeneration.

Discussion and conclusions

Evaluation of study approaches

In this study, we used several climate change projections, with wide variations in temperatures and precipitation, in simulations by forest ecosystem (SIMA) and mechanistic wind damage (HWIND) models, in order to study the changes to regional risks of wind damage in Finnish upland boreal forests under different tree species preferences in forest regeneration over the period 2010–2099. Following the approach of Ikonen et al. (2017), we calculated the annual probability for wind damage and the amount of wind damage, for NFI plots throughout Finland, based on the predicted CWSs. Compared to our study, recent studies on climate change impacts on forests have mainly considered either the effects of climate change on productivity or on disturbances (Reyer et al. 2017). This is, although disturbances may affect forest productivity, e.g., via a decrease in



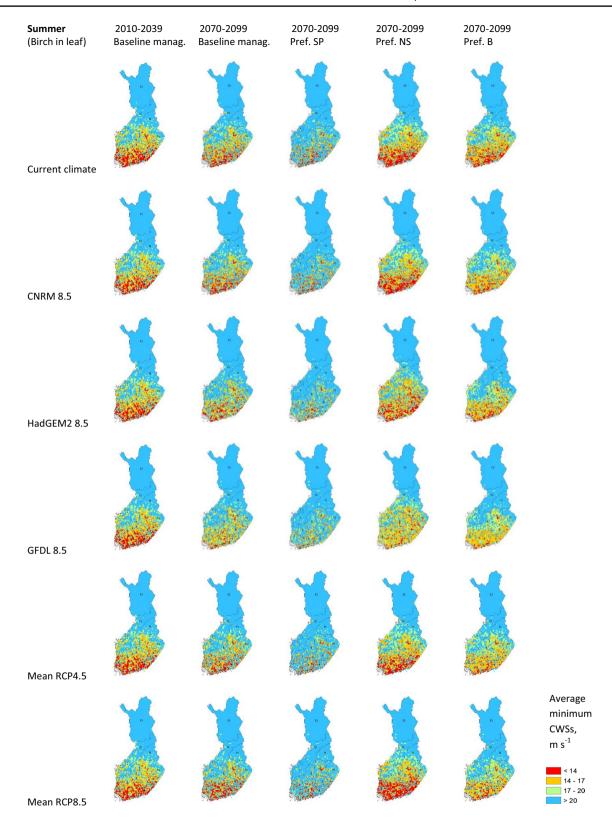


Fig. 3 Average minimum CWSs (m $\rm s^{-1}$) in summer (birch in leaf) in the periods 2010–2039 and 2070–2099 under each climate change projection and management scenario. Red color indicates high wind damage risk. Medium-fertility sites were planted with Scots pine

(Pref. SP), Norway spruce (Pref. NS) or silver birch (Pref. B), or with the tree species that was dominant before the final clear-felling (baseline management). (Color figure online)



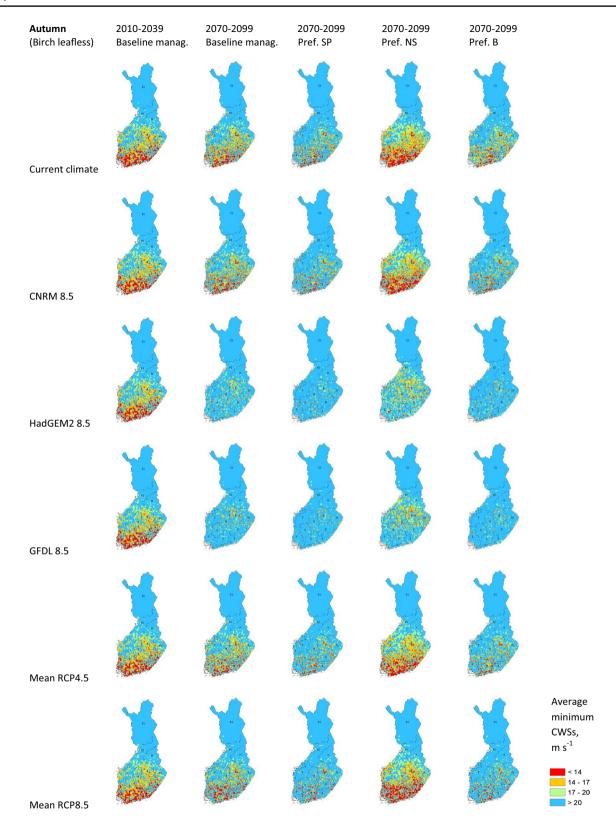


Fig. 4 Average minimum CWSs (m $\rm s^{-1}$) in autumn (birch leafless) in the periods 2010–2039 and 2070–2099 under each climate change projection and management scenario. Red color indicates high wind damage risk. Medium-fertility sites were planted with Scots pine

(Pref. SP), Norway spruce (Pref. NS) or silver birch (Pref. B), or with the tree species that was dominant before the final clear-felling (baseline management). (Color figure online)



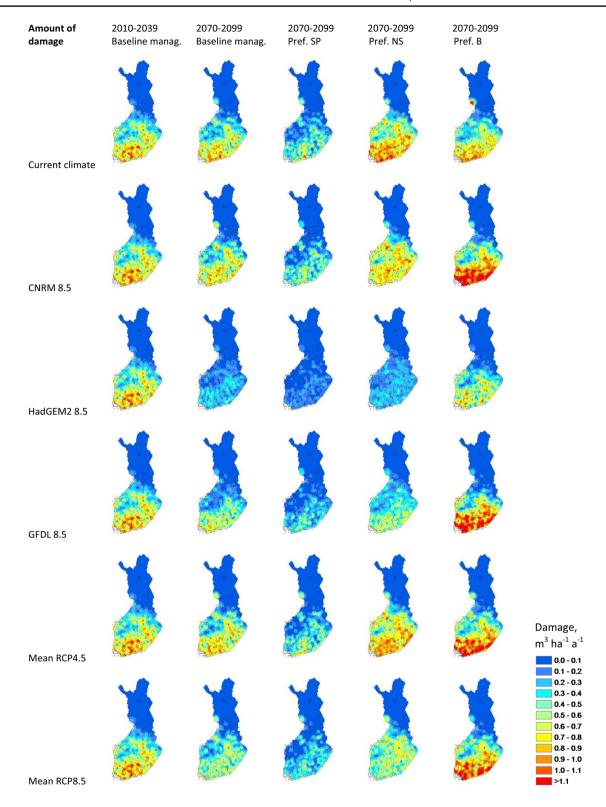


Fig. 5 Predicted amount of damage (m³ ha⁻¹ a⁻¹) in summer in the periods 2010–2039 and 2070–2099 under each climate change projection and management scenario. Red color indicates high wind damage risk. Medium-fertility sites were planted with Scots pine

(Pref. SP), Norway spruce (Pref. NS) or silver birch (Pref. B), or with the tree species that was dominant before the final clear-felling (baseline management). (Color figure online)



growing stock. Furthermore, the susceptibility to a certain disturbance is also affected by the forest structure (age, species) as controlled by forest management, and development phase of the forest.

On the other hand, we used outputs of the SIMA model simulations for 2010–2099 as inputs for HWIND, which predicted the CWSs needed to cause damage to various stand configurations. In addition, calculation of the risk of wind damage to the plots was done by always assuming new and vulnerable upwind edges, and considering only the most vulnerable tree cohort of a plot. As a result, we may have overestimated the predicted amount of wind damage to some degree. Furthermore, because damaged trees were not excluded from the stands during the SIMA model simulations, we may also have overestimated the forest growth in stands suffering damage, at least to some degree, over time. In future studies, damaged trees should be removed annually from the stands when simulating forest growth and dynamics over time.

The use of ecosystem modeling as combined with mechanistic wind damage modeling allowed us to study the effects of the growth responses of different tree species, forest dynamics and tree species preferences on the risk of wind damage under different climate projections. This would not be possible using statistical growth models alone, assuming no change in climate over time. The growth responses of the trees were also controlled in our simulations by water and nitrogen availability, which were affected by edaphic factors (climate and site). This allowed us to evaluate also how we should adapt the use of different tree species in forest regeneration in different regions, in order to better consider the uncertainties related to climate change and its projected impacts on forestry.

Evaluation of main findings

Based on our study, the use of individual GCMs, especially such as HadGEM2 8.5 and GFDL 8.5, affected more the development of a proportion of the tree species and the volume of growing stock (Fig. 1, Appendix Tables 6, 7), and thus the amount of damage (Fig. 2, Appendix Table 5), compared to the multi-model mean GCM projections (mean RCP4.5 and mean RCP8.5) used by Ikonen et al. (2017). Use of the most extreme GCM projections also affected the predicted amount of wind damage more than the increased use of different tree species in forest regeneration in the long term. When the GCMs with smaller changes in temperature and precipitation, such as

MPI 4.5 and MPI 8.5, were used, our results were similar to those of Ikonen et al. (2017).

Compared to the current climate, under CNRM 8.5 (significant increase in temperature, greatest increase in precipitation), with an increased use of birch in southern Finland, the predicted amount of wind damage (along with forest growth and volume of growing stock) increased markedly. On the other hand, under strong climate warming and decreased precipitation compared to the current climate, the growing conditions became unfavorable, especially for Norway spruce, as shown previously by Alrahahleh et al. (2018). Thus, especially under HadGEM2 8.5 in southern and central Finland, the volume of growing stock and the predicted amount of damage were remarkably low. In line with previous studies (e.g., Ikonen et al. 2017), the increased use of Scots pine in forest regeneration seemed to decrease the wind damage risks, as opposed to an increased use of Norway spruce. The trend of decreasing wind damage risk with increasing proportion of birch was, however, quite clear for autumn, when the birch was leafless. Norway spruce has been found the most vulnerable conifer to wind damage also in wind storms in both northern and central Europe (Schmidt et al. 2010; Reyer et al. 2017).

The growth (and mortality) responses of different tree species varied in regard to climate change severity, which greatly affected the proportions of tree species (of volume of growing stock) at the regional level as was reported by Alrahahleh et al. (2018). For example, with an increased use of Norway spruce, its proportion ranged from 1% (GFDL 8.5) to 64% (MPI 4.5; Appendix Table 6). GFDL 8.5 predicted a very large increase in temperature, but only a moderate increase in precipitation in southern Finland, causing unfavorable conditions for Norway spruce. Contrarily, MPI 4.5 only produced a minor increase in temperature, but, as a result, the proportion of Norway spruce was the greatest under MPI 4.5, compared to the other climate change projections. Thus, the use of different tree species in forest regeneration will affect their proportions more under mild climate change than under severe climate change.

The most important factor for CWSs, and the probability of wind damage at the regional level, was in our study, the proportion of different tree species. At the tree and stand level, the wind damage risk and CWSs are largely affected by the different tree and stand characteristics and stand configurations (Peltola et al. 1999b; Zubizarreta-Gerendiain et al. 2012; Dupont et al. 2015; Suvanto et al. 2016). In



our study, climate change directly affected the growth and mortality of the trees and, indirectly, the CWSs and the probability of damage, which in turn, together with the volume of the growing stock, affected the amount of wind damage.

Under Finnish conditions, relatively low wind speeds (i.e., CWSs $< 17 \text{ m s}^{-1}$) have caused wind damage in recent decades (Laiho 1987; Gregow et al. 2011a, 2011b; Zubizarreta-Gerendiain et al. 2012). Since 2000, about 1.6 million m³ a⁻¹ of timber has been damaged by windstorms, mostly in southern and central Finland (Laiho 1987; Gregow et al. 2011a; Zubizarreta-Gerendiain et al. 2012). By comparison, if we assume an annual regeneration area of about 1-2% (Finnish Forest Research Institute 2014), and that each new upwind stand edge will be vulnerable about 20 years, according to our calculations, the total damaged amount of timber throughout Finland would be, on average, 0.5-2.4 and 0.9-2.6 million m³ a⁻¹ under different climate projections, with an increased planting of Scots pine and baseline management. In contrast, with an increased planting of Norway spruce or birch, the corresponding ranges would be higher, 1.0-3.0 and 1.9–3.0 million m³ a⁻¹, on average, respectively. The corresponding ranges calculated by Ikonen et al. (2017), based on the current climate and multi-model mean GCM projections, were of the same magnitude, but the ranges were narrower.

In the long term, the increasing risks to forests may, at least partly, counteract the expected increase in forest productivity under changing climate (Kellomäki et al. 2008; Reyer et al. 2017; Seidl et al. 2017). The higher share of Norway spruce might result in increased wind damage to forests (Schmidt et al. 2010; Reyer et al. 2017). It might cause also a greater risk of biotic damage (e.g., by wood decay and bark beetles; Subramanian et al. 2016; Thom and Seidl 2016; Honkaniemi et al. 2017), which were not considered in this study. Increasing abiotic and biotic damage risks to forests should be considered in adapting forest management strategies to properly accommodate/counteract projected climate change (Peltola et al. 2010; Seidl et al. 2011; Hanewinkel et al. 2013; Subramanian et al. 2016; Reyer et al. 2017). For example, growing forests with more climate change adapted tree species (genotypes), and their mixtures, could help to reduce the possible negative effects of climate change on forests (Neuner et al. 2015; Metz et al. 2016; Anyomi et al. 2017; Jactel et al. 2017). The wind damage risk to forests could also be lessened by avoiding the creation of large height differences among adjacent older stands and using shorter rotation lengths (Zeng et al. 2007; Heinonen et al. 2009; Jactel et al. 2009; Zubizarreta-Gerendiain et al. 2012).



The increasing use of certain tree species in forest regeneration may decrease the risk of wind damage in the long term, but the effect will depend on the current structure of the forests (age, tree species proportions), geographical region (climate and site) and the severity of the climate change, as well as the season and applied forest management intensity. Thus, opposite measures for risk management could be suggested, dependent on which climate change projection is used and the time span studied. In this sense, it is important to understand how different climate projections and management choices can affect the wind damage risk over the short and long term. This is crucial in order to identify the most appropriate and most flexible adaptation measures to climate change.

In future studies, different wood harvesting scenarios should be considered, as the intensity of wood harvesting can greatly affect the development of forest resources and, consequently, the wind damage risks. In addition, the effects of different abiotic and biotic damage risks to forests should be considered. More integrated models, with the capability of removing damaged trees from the forest during simulations, should be developed. All in all, the importance of considering the risks of various disturbances is increasing in forest management because such conditions can amplify, or even cancel out, any expected increases in forest productivity. Although our study represented a Finnish boreal case study, our findings could be applicable also to other regions especially in Northern Europe, but also partially in central Europe.

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Appendix

See Tables 3, 4, 5, 6 and 7.

Table 3 Average minimum critical wind speeds (m s⁻¹) in summer (birch in leaf)

	1	2	3	4	5	6	7	8	9	10	11	12	13
	CU	H45	H85	M45	M85	Ca45	Ca85	Mi45	Mi85	CN85	GF85	R45	R85
Southern Fin	land												
2010-2039	a												
Baseline	17	17	17	17	17	17	17	17	17	17	17	17	17
2040-2069													
Baseline	18	18	18	18	18	18	18	18	18	18	18	18	18
Pref. SP	19	20	19	20	20	20	20	20	19	20	19	20	20
Pref. NS	17	17	17	17	17	17	17	17	17	17	16	17	17
Pref. B	17	18	17	17	18	17	17	17	18	18	17	17	18
2070-2099													
Baseline	18	18	18	18	18	18	18	19	18	18	18	18	18
Pref. SP	21	21	20	21	21	21	21	21	21	21	21	21	21
Pref. NS	16	16	16	17	17	17	16	17	16	16	17	17	16
Pref. B	17	18	17	17	17	18	17	18	17	17	18	17	17
Central Finla	nd												
2010-2039	a												
Baseline	19	19	19	19	19	19	19	19	19	20	19	19	19
2040-2069													
Baseline	20	21	21	21	21	21	21	21	21	21	21	21	21
Pref. SP	21	22	22	22	22	22	22	22	22	22	22	22	22
Pref. NS	20	20	20	20	20	20	20	20	20	20	20	20	20
Pref. B	20	20	20	20	20	20	20	20	20	20	20	20	20
2070-2099													
Baseline	21	21	21	21	21	21	21	21	21	21	21	21	21
Pref. SP	23	23	23	23	23	23	23	23	23	23	23	23	23
Pref. NS	19	19	19	19	19	19	18	19	19	19	19	19	19
Pref. B	20	20	20	20	20	20	20	20	20	20	20	20	20
Northern Fin	land												
2010-2039	a												
Baseline	27	27	27	27	27	28	28	27	28	27	28	27	28
2040-2069													
Baseline	28	29	29	29	29	29	29	29	29	29	29	29	29
Pref. SP	28	29	29	29	29	29	29	29	29	29	29	29	29
Pref. NS	28	28	28	28	28	28	28	28	28	28	28	28	28
Pref. B	28	29	29	28	28	29	29	29	29	29	29	29	29
2070-2099													
Baseline	28	29	28	29	29	29	29	29	29	29	29	29	29
Pref. SP	29	30	29	30	30	30	30	30	30	30	30	30	30
Pref. NS	28	28	27	28	28	28	27	28	27	28	27	28	28
Pref. B	28	28	28	28	28	28	28	28	28	28	28	28	28

Climate change projections: 1=current climate, 2=HadGEM2 4.5, 3=HadGEM2 8.5, 4=MPI 4.5, 5=MPI 8.5, 6=CanESM2 4.5, 7=CanESM2 8.5, 8=MIROC5 4.5, 9=MIROC5 8.5, 10=CNRM 8.5, 11=GFDL 8.5, 12=Mean RCP4.5, 13=Mean RCP8.5



^aNo remarkable differences between management scenarios in the first period

Table 4 Probabilities for wind damage (dimensionless, in a range of 0–1)

Note													- 10	
Southern Final		1	2	3	4	5 M95	6	7	8	9 M:05	10	11 CE95	12 D45	13
Baseline 0.8		CU	H45	H85	W145	M85	Ca45	Cass	M145	W1185	CN85	GF85	K45	K85
Baseline 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 2040-2069	Southern Fin	land												
Baseline	2010–2039	a												
Baseline 0.7 0.5	Baseline	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8
Pref. SP	2040–2069													
Pref. NS	Baseline	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7
Pref. B 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 Pref. SP 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 Pref. NS 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 Pref. B 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 Pref. B 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 Pref. B 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 Pref. B 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 Pref. SP 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 Pref. SP 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 Pref. B 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 Pref. B 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 Pref. SP 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 Pref. SP 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 Pref. SP 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 Pref. SP 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 Pref. SP 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 Pref. SP 0.3 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 Pref. SP 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 Pref. SP 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 Pref. SP 0.0 0.0 0.0 0.0	Pref. SP	0.6	0.5	0.5	0.5	0.5	0.6	0.6	0.6	0.6	0.5	0.6	0.5	0.5
Baseline 0.7 0.6 0.7 0.6 0.6 0.6 0.7 0.6 0.7 0.6 0.6 0.7 0.6 0.6 0.7 0.6 0.6 0.6 0.7 0.6 0.6 0.7 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.8	Pref. NS	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8
Baseline 0.7 0.6 0.7 0.6 0.6 0.6 0.7 0.6 0.7 0.6 0.7 0.6 0.6 Pref. SP 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 Pref. NS 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 Pref. B 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 Pref. B 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 Pref. B 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 Pref. B 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 Pref. B 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 Pref. SP 0.8 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 Pref. SP 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 Pref. NS 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 Pref. B 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 0.6 Pref. SP 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 Pref. NS 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.7 Pref. B 0.6 0.6 0.5 0.5 0.6 0.6 0.5 0.5 0.5 0.5 0.5 0.5 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 Pref. NS 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 Pref. SP 0.1	Pref. B	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8
Pref. SP	2070-2099													
Pref. NS 0.8 0.	Baseline	0.7	0.6	0.7	0.6	0.6	0.6	0.7	0.6	0.7	0.6	0.7	0.6	0.6
Pref. B	Pref. SP	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4
Central Final	Pref. NS	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8
Baseline 0.6	Pref. B	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8	0.8
Baseline O.5 O.5														
Baseline 0.5 0.6 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.	Baseline	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6
Pref. SP 0.4 0.6 0.5 0.	2040-2069													
Pref. NS 0.6 0.5 0.6 0.5 0.	Baseline	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5
Pref. B 0.6 0.5 0.6 0.5 0.6 0.7	Pref. SP	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4
Baseline 0.5	Pref. NS	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6
Baseline 0.5 0.3 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.7 0.	Pref. B	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6	0.6
Pref. SP 0.3 0.7 0.1 0.1 0.1 0.	2070-2099													
Pref. NS 0.7 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.	Baseline	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5
Pref. B 0.6 0.6 0.5 0.6 0.6 0.6 0.5 0.6 0.5 0.6 0.5 0.6 0.1	Pref. SP	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3
Northern Finland 2010–2039 ³ Baseline 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1	Pref. NS	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7
2010–2039 ^a Baseline 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1	Pref. B	0.6	0.6	0.5	0.6	0.6	0.6	0.5	0.6	0.5	0.6	0.5	0.6	0.6
Baseline 0.1 0.	Northern Fin	land												
2040–2069 Baseline 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1	2010-2039	a												
Baseline 0.1 0.0 0.	Baseline	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
Pref. SP 0.1 0.0 0.	2040-2069													
Pref. SP 0.1 0.0 0.	Baseline	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
Pref. B 0.1 0.	Pref. SP	0.1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Pref. B 0.1 0.	Pref. NS	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
2070–2099 Baseline 0.1 0.0 Pref. SP 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0														
Baseline 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1														
Pref. SP 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.			0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
1101.1.0 0.1 0.1 0.1 0.1 0.1 0.1 0.1 0.1	Pref. NS	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
Pref. B 0.1 0.1 0.0 0.1 0.1 0.1 0.0 0.1 0.1 0.1														

Climate change projections: **1**=current climate, **2**=HadGEM2 4.5, **3**=HadGEM2 8.5, **4**=MPI 4.5, **5**=MPI 8.5, **6**=CanESM2 4.5, **7**=CanESM2 8.5, **8**=MIROC5 4.5, **9**=MIROC5 8.5, **10**=CNRM 8.5, **11**=GFDL 8.5, **12**=Mean RCP4.5, **13**=Mean RCP8.5



^aNo remarkable differences between management scenarios in the first period

Table 5 Predicted amount of damage (m³ ha⁻¹ a⁻¹)

	1	2	3	4	5	6	7	8	9	10	11	12	13
	CU	2 Н45	э Н85	4 M45	M85	Ca45	Ca85	o Mi45	Mi85	CN85	GF85	R45	R85
Coutharn Ein	land												
Southern Fin 2010–2039													
Baseline	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7	0.7
2040-2069													
Baseline	0.5	0.5	0.5	0.6	0.6	0.6	0.5	0.5	0.5	0.6	0.5	0.6	0.6
Pref. SP	0.4	0.4	0.4	0.5	0.4	0.5	0.4	0.5	0.4	0.5	0.4	0.5	0.4
Pref. NS	0.6	0.5	0.5	0.6	0.6	0.6	0.5	0.6	0.5	0.6	0.5	0.6	0.6
Pref. B	0.6	0.7	0.7	0.7	0.7	0.8	0.7	0.8	0.8	0.8	0.8	0.7	0.8
2070-2099													
Baseline	0.6	0.6	0.3	0.7	0.6	0.6	0.4	0.6	0.5	0.6	0.5	0.6	0.5
Pref. SP	0.4	0.3	0.1	0.4	0.4	0.4	0.2	0.4	0.3	0.4	0.3	0.4	0.3
Pref. NS	0.8	0.6	0.2	0.8	0.8	0.7	0.3	0.7	0.4	0.7	0.5	0.8	0.5
Pref. B	0.7	0.9	0.6	0.8	0.9	0.9	0.8	0.9	0.9	1.0	0.9	0.9	0.9
Central Finla	nd												
2010-2039	a												
Baseline	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5	0.5
2040-2069													
Baseline	0.4	0.4	0.3	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4
Pref. SP	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3
Pref. NS	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.4	0.5	0.4	0.4	0.4
Pref. B	0.4	0.4	0.4	0.4	0.4	0.5	0.4	0.5	0.4	0.5	0.5	0.5	0.5
2070-2099													
Baseline	0.4	0.4	0.2	0.5	0.5	0.4	0.3	0.4	0.3	0.5	0.3	0.5	0.4
Pref. SP	0.3	0.2	0.1	0.3	0.3	0.3	0.1	0.3	0.2	0.3	0.2	0.3	0.2
Pref. NS	0.6	0.5	0.2	0.6	0.7	0.6	0.3	0.6	0.3	0.6	0.3	0.6	0.5
Pref. B	0.4	0.5	0.4	0.5	0.5	0.5	0.5	0.5	0.5	0.6	0.5	0.5	0.5
Northern Fin	land												
2010-2039	a												
Baseline	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
2040-2069													
Baseline	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
Pref. SP	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
Pref. NS	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
Pref. B	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
2070-2099													
Baseline	0.1	0.1	0.0	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.0	0.1	0.1
Pref. SP	0.1	0.1	0.0	0.1	0.1	0.1	0.0	0.1	0.0	0.1	0.0	0.1	0.1
Pref. NS	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
Pref. B	0.1	0.1	0.0	0.1	0.1	0.1	0.0	0.1	0.1	0.1	0.0	0.1	0.1

Climate change projections: **1**=current climate, **2**=HadGEM2 4.5, **3**=HadGEM2 8.5, **4**=MPI 4.5, **5**=MPI 8.5, **6**=CanESM2 4.5, **7**=CanESM2 8.5, **8**=MIROC5 4.5, **9**=MIROC5 8.5, **10**=CNRM 8.5, **11**=GFDL 8.5, **12**=Mean RCP4.5, **13**=Mean RCP8.5



^aNo remarkable differences between management scenarios in the first period

Table 6 The proportion of each tree species (percentage of stem volume, %) under each climate change projection in the first period 2010–2039 (P1) under baseline management and in the third period 2070–2099 (P3) under each management scenario

	Clin	nate ch	ange pi	ojectio	n	-	-	,					
	1	2	3	4	5	6	7	8	9	10	11	12	13
	CU	H45	H85	M45	M85	Ca45	Ca85	Mi45	Mi85	CN85	GF85	R45	R85
Scots pine													
Baseline													
P1													
North	65	65	65	65	65	65	65	65	65	65	65	65	65
Central	51	51	51	51	51	51	51	51	51	51	51	51	51
South	42	42	42	42	42	42	42	42	42	42	42	42	42
Baseline	72	72	72	72	72	72	72	72	72	72	72	72	72
P3													
North	62	69	75	67	68	70	75	70	73	69	75	68	70
Central	52	62	73 74	59	61	61	73 72	61	73 71	61	69	60	64
South	48	56	62	52	54	54	63	55	59	56	54	53	61
Pref. SP	40	30	02	32	34	34	03	33	39	30	34	33	01
P3	75	00	0.5	70	70	70	0.4	70	02	70	0.4	70	00
North	75 72	80	85	78	79 70	79 70	84	79 70	83	79 70	84	79 70	80
Central	73	79	89	77	79	78	87	79	85	78	83	78	81
South	72	78	87	75	77	76	85	76	81	77	77	75	81
Pref. NS													
P3													
North	47	52	57	50	51	53	58	53	56	52	59	51	53
Central	33	39	52	36	38	39	53	40	52	39	52	37	41
South	27	34	40	29	30	32	42	33	41	34	41	30	40
Pref. B													
P3													
North	52	61	62	57	59	62	63	62	64	61	63	59	62
Central	41	42	31	50	51	44	36	43	36	47	34	48	44
South	38	28	20	40	37	29	22	28	22	34	20	34	30
Norway spru	ice												
Baseline													
P1													
North	22	21	22	22	22	21	21	21	21	22	21	21	22
Central	31	30	30	31	31	30	30	30	30	31	30	31	31
South	40	38	38	39	39	38	38	38	38	39	37	39	38
Baseline													
P3													
North	29	24	19	26	25	24	19	23	20	24	17	25	23
Central	39	29	6	36	34	30	8	30	10	32	5	34	27
South	43	22	1	39	35	27	2	24	3	29	1	35	18
Pref. SP													
P3													
North	20	16	11	17	17	16	12	16	13	17	11	17	16
Central	22	14	2	19	18	16	3	15	5	17	2	18	13
South	23	9	1	19	16	13	1	11	1	14	1	17	7
Pref. NS	-	-			-							-	-
P3													
North	47	43	38	46	45	43	38	43	39	43	35	44	43
Central	62	53	15	60	57	54	20	52	24	55	10	57	50
South	67	44	3	64	61	52	6	47	7	53	1	61	38
	07			04	01	J2		7/		55	1	01	



 Table 6 (continued)

	Clin	nate ch	ange pr	ojectio	n								
	1	2	3	4	5	6	7	8	9	10	11	12	13
	CU	H45	H85	M45	M85	Ca45	Ca85	Mi45	Mi85	CN85	GF85	R45	R85
Pref. B													
P3													
North	21	18	12	20	20	18	13	18	15	19	12	19	18
Central	27	15	2	23	20	17	3	16	4	19	2	20	13
South	27	8	1	22	17	12	1	10	1	13	1	17	7
Birch													
Baseline													
P1													
North	13	13	13	13	13	13	14	14	14	13	14	14	13
Central	18	19	19	18	18	19	19	19	19	19	19	19	19
South	18	19	19	19	19	19	20	20	20	19	20	19	19
Baseline													
P3													
North	9	7	7	7	6	7	7	7	7	7	8	7	6
Central	8	9	21	6	6	9	19	9	19	7	27	7	9
South	9	22	36	9	11	18	35	21	38	15	45	11	21
Pref. SP													
P3													
North	6	4	4	4	4	4	4	5	4	5	5	4	4
Central	5	6	9	4	4	6	10	6	10	5	14	4	6
South	6	13	13	6	7	11	15	13	18	9	23	8	11
Pref. NS													
P3													
North	5	4	5	4	4	5	5	5	5	4	6	5	4
Central	5	8	34	4	5	7	27	8	24	6	37	5	8
South	6	22	57	6	9	16	52	20	52	13	58	9	22
Pref. B													
P3													
North	27	21	26	23	21	20	24	19	22	20	25	22	21
Central	32	42	66	27	29	39	61	41	59	34	64	32	42
South	35	63	80	39	46	59	77	61	77	53	80	48	63

 $\begin{array}{l} \textbf{1} = \text{current climate, } \textbf{2} = \text{HadGEM2 4.5, } \textbf{3} = \text{HadGEM2 8.5, } \textbf{4} = \text{MPI 4.5, } \textbf{5} = \text{MPI 8.5, } \textbf{6} = \text{CanESM2 4.5, } \\ \textbf{7} = \text{CanESM2 8.5, } \textbf{8} = \text{MIROC5 4.5, } \textbf{9} = \text{MIROC5 8.5, } \textbf{10} = \text{CNRM 8.5, } \textbf{11} = \text{GFDL 8.5, } \textbf{12} = \text{Mean RCP4.5, } \textbf{13} = \text{Mean RCP8.5} \\ \end{array}$



Table 7 Average volume of growing stock (m³ ha⁻¹)

	1	2	3	4	5	6	7	8	9	10	11	12	13
	CU	H45	H85	M45	M85	Ca45	Ca85	Mi45	Mi85	CN85	GF85	R45	R85
Southern Fin	land												
2010-2039	ì												
Baseline	129	132	132	133	133	133	133	132	132	133	133	133	134
2040-2069													
Baseline	109	111	101	120	121	117	109	115	110	123	108	119	119
Pref. SP	116	118	108	125	124	122	115	122	115	129	113	126	124
Pref. NS	110	100	86	113	112	105	95	104	94	115	91	110	108
Pref. B	114	123	119	123	125	129	126	129	127	133	129	128	130
2070-2099													
Baseline	140	126	64	152	147	140	86	138	101	144	107	148	119
Pref. SP	145	134	71	152	150	144	97	143	109	150	108	150	130
Pref. NS	146	109	44	151	142	127	61	123	73	130	80	143	95
Pref. B	128	142	89	140	146	151	117	152	132	157	139	149	141
Central Finla 2010–2039													
Baseline	122	125	126	125	126	126	127	126	126	127	128	126	126
2040-2069													
Baseline	107	115	108	118	120	119	114	117	114	123	112	120	120
Pref. SP	113	119	114	122	123	123	120	122	119	127	119	123	126
Pref. NS	108	107	98	114	114	111	103	110	104	117	100	114	113
Pref. B	110	118	116	118	119	122	121	124	122	125	124	121	123
2070-2099													
Baseline	129	139	78	148	151	146	96	147	106	154	108	149	136
Pref. SP	134	141	90	148	151	147	108	148	117	155	117	150	142
Pref. NS	134	132	54	148	149	142	71	141	83	150	84	146	126
Pref. B	118	136	98	129	136	141	116	141	126	146	130	135	137
Northern Fin 2010–2039													
Baseline	94	100	101	99	99	102	102	102	102	100	103	100	101
2040-2069													
Baseline	102	130	133	124	128	134	136	134	137	134	140	129	135
Pref. SP	104	130	135	123	127	134	138	135	138	135	141	130	136
Pref. NS	104	124	127	120	124	129	130	129	131	129	133	125	130
Pref. B	101	124	129	117	121	128	132	128	132	129	137	123	129
2070-2099													
Baseline	126	168	142	161	173	174	153	172	162	181	161	169	175
Pref. SP	128	169	144	161	172	172	155	173	161	180	163	170	176
Pref. NS	130	166	133	161	171	172	146	173	157	179	153	168	173
Pref. B	120	154	135	148	157	158	147	158	153	167	155	155	161

Climate change projections: **1**=current climate, **2**=HadGEM2 4.5, **3**=HadGEM2 8.5, **4**=MPI 4.5, **5**=MPI 8.5, **6**=CanESM2 4.5, **7**=CanESM2 8.5, **8**=MIROC5 4.5, **9**=MIROC5 8.5, **10**=CNRM 8.5, **11**=GFDL 8.5, **12**=Mean RCP4.5, **13**=Mean RCP8.5

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