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#### **ORIGINAL PAPER**



# Long-term effects of fire and harvest on carbon stocks of boreal forests in northeastern China

Chao Huang<sup>1,2</sup> • Hong S. He<sup>3,4</sup> • Yu Liang<sup>1</sup> • Zhiwei Wu<sup>1</sup> • Todd J. Hawbaker<sup>5</sup> • Peng Gong<sup>6</sup> • Zhiliang Zhu<sup>7</sup>

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#### Abstract

• *Key message* Fire, harvest, and their spatial interactions are likely to affect boreal forest carbon stocks. Repeated disturbances associated with short fire return intervals and harvest rotations resulted in landscapes with a higher proportion of young stands that store less carbon than mature stands.

• *Context* Boreal forests represent about one third of forest area and one third of forest carbon stocks on the Earth. Carbon stocks of boreal forests are sensitive to climate change, natural disturbances, and human activities.

• *Aims* The objectives of this study were to evaluate the effects of fire, harvest, and their spatial interactions on boreal forest carbon stocks of northeastern China.

• *Methods* We used a coupled forest landscape model (LANDIS PRO) and a forest ecosystem model (LINKAGES) framework to simulate the landscape-level effects of fire, harvest, and their spatial interactions over 150 years.

• *Results* Our simulation suggested that aboveground carbon and soil organic carbon are significantly reduced by fire and harvest over the whole simulation period. The long-term effects of fire and harvest on carbon stocks were greater than the short-term effects. The combined effects of fire and harvest on carbon stocks are less than the sum of the separate effects of fire and harvest. The response of carbon stocks was impacted by the spatial variability of fire and harvest regimes.

• *Conclusion* These results emphasize that the spatial interactions of fire and harvest play an important role in regulating boreal forest carbon stocks.

Keywords Fire · Harvest · Carbon stocks · LANDIS PRO · LINKAGES · Model coupling

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**Contribution of the co-authors** Designed figures and tables in the manuscript: Chao Huang, Hong S. He, Zhiwei Wu, and Yu Liang. Conceived and designed the experiments: Chao Huang, Hong S. He, Yu Liang, and Zhiwei Wu. Analyzed the data: Chao Huang, Yu Liang, Hong S. He, and Todd J. Hawbaker. Wrote and revise the paper: Chao Huang, Hong S. He, Yu Liang, Todd J. Hawbaker, Peng Gong, and Zhiliang Zhu.

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## **1** Introduction

Boreal forests are an important component of the global carbon cycle (Lutz et al. 2013). They store  $52.7 \pm 1.7$  Pg carbon in above ground biomass and  $170.5 \pm 3.8$  Pg carbon in soil (Pan et al. 2011). The carbon stocks of boreal forests are strongly affected by fire and harvest (Barker et al. 2014). Fire and harvest alter forest productivity, release carbon directly into the atmosphere, and transfer carbon from live biomass into detritus, soils, and forest products (Bhatti et al. 2002; Kasischke et al. 2013). The carbon balance of boreal forests depends on the frequency and intensity of these disturbances (Chen et al. 2003; Clarke et al. 2015). For example, boreal forests in central and eastern Canada were carbon sources ( $29 \pm 21$  g m<sup>-2</sup> C year) between 1990 and 1998 due to high-intensity fire, harvest, and insect outbreaks (Chen et al. 2003). Furthermore, climate change may increase the frequency and intensity of fire and insect disturbances, which will further exacerbate the loss of carbon from boreal forests (O'Donnell et al. 2011). Therefore, evaluating and distinguishing the effects of fire and harvest on boreal forest carbon stocks is of great importance for predicting the response of boreal forest carbon dynamics to climate change.

Boreal forests in the Great Xing' an Mountains cover  $8.46 \times$  $10^5$  km<sup>2</sup>, store about 350 Tg aboveground carbon, and play an important role in maintaining China's carbon balance (Fu et al. 2013). However, boreal forests in this region are facing pressures from repeated fires and timber harvesting activities. A total of 1614 fires burned  $3.52 \times 10^6$  ha of boreal forests and released 293 Tg carbon into the atmosphere during the 46-year time period from 1965 to 2010 in the Great Xing'an Mountains (Hu et al. 2012). Meanwhile,  $3.38 \times 10^6$  ha of boreal forests has been harvested since the 1960s in this area (Zhang et al. 2009). Timber harvesting in this region is primarily clear-cutting, and these harvest activities have greatly altered forest composition, ecosystem processes, and landscape pattern. Furthermore, postfire salvage logging is a common forest management practice in the Great Xing'an Mountains (Chen et al. 2014), with the potential to alter forest regeneration and carbon stocks (Serrano-Ortiz et al. 2011). In order to maintain the productivity and ecological functions of China's boreal forests, it is essential to assess the long-term effects of repeated fire and harvest and their spatial interactions on carbon stocks.

Field studies have investigated the effects of fire and harvest in boreal forests on aboveground biomass (Wirth et al. 1999), soil organic carbon (Hazlett et al. 2007), and forest floor carbon (Boby et al. 2010; Shrestha and Chen 2010). These studies demonstrated that the effects of fire and harvest on boreal forest carbon stocks are a function of the balance between carbon loss from those disturbances and carbon gain from vegetation recovery. Normally, vegetation recovery after disturbance will transition forest stands from carbon sources to carbon sinks, and with sufficient time, the forest will sequester all the carbon lost from disturbances (Kashian et al. 2006). This implies that the



long-term effects of fire and harvest on forest ecosystem carbon stocks are smaller than their short-term effects. However, most field studies were carried out using a space-for-time substitution within a relatively short time span (within a few decades) that was limited in capturing the inherently variable and stochastic nature of disturbances and vegetation recovery over long time periods (Piirainen et al. 2015). Although the space-for-time substitution approach can overcome limitations of short-duration field studies, they may fail to capture longer-term ecological processes of postdisturbance recovery (Gough et al. 2007). In addition, because climate change is likely to alter the frequency and severity of fire and species composition in boreal forests, the pathway of forest recovery and carbon stocks may also be affected (Anderson-Teixeira et al. 2013; Kasischke and Stocks 2012).

Ecological models are effective tools for studying the effects of repeated fire and harvest on forest carbon dynamics over long terms (Chertov et al. 2009). They have been used to investigate the effects of repeated fire and harvest on boreal forest carbon stocks in North America (Goetz et al. 2012; Govind et al. 2011), Europe (Moroni et al. 2010; Shanin et al. 2011), and China (Jiang et al. 2002; Luo et al. 2014). Additionally, ecological models can evaluate the uncertainties in how fire and harvest affect boreal forest carbon stocks by running replicate simulations (Kurz et al. 2008). Uncertainties in projections because of the stochastic nature of disturbances can be assessed by running multiple simulation scenarios. Moreover, separating the effects of fire and harvest has practical value for forest management policy-making that aims to minimize carbon loss from natural disturbances and human activities. However, few model-based studies have quantified the relative importance of periodic disturbances on boreal forest carbon stocks (Gustafson et al. 2010; Li et al. 2013).

The objective of this study was to quantify the long-term effects and relative importance of fire and harvest on carbon stocks of boreal forests in the Great Xing'an Mountains northeastern China. To do this, we used a coupled forest ecosystem and landscape model framework to predict aboveground carbon and soil organic carbon stocks at the landscape level. Specifically, we (1) examined the dynamics of aboveground carbon and soil organic carbon, as they changed in response to repeated fire and harvest; (2) quantified the relative importance of fire and harvest on boreal forest carbon stocks; and (3) assessed the effects of the spatial patterns of fire and harvest and their interactions on carbon stocks over 150 years.

# 2 Methods

## 2.1 Study area

Our study area included the Huzhong Natural Reserve (166,906 ha) and the Huzhong Forestry Bureau (770,338 ha), which are located in the northern side of the

Great Xing'an Mountains (Fig. 1). The elevation ranges from 406 m in the northeast to 1514 m in the southwest. The climate is cool continental climate, with annual mean temperature of -4.7 °C and annual precipitation of about 497 mm.

Our study area is classified into 26 ecoregions based on topography, forest types, and soil types (Fig. 1, Table S1). The terrace ecoregions (ecoregions 1–8) contain 7% of the forests in our study area and are dominated by dahurian larch (*Larix gmelini* (Rupr. Kuzen)), white birch (*Betula platyphylla* Suk), poplar (*Populus suaveolens* Fisch), and willow (*Chosenia arbutifolia* (Pall.) A. Skv). Soils are sandy loam textured and poorly drained and have relatively low cation exchange capacity. Additionally, the Huma River and majority of the roads in the study area are located in the terrace ecoregions.

The south-slope ecoregions (ecoregions 9–16) include 36% of the forests in this region and are dominated by dahurian larch, white birch, aspen (*Populus davidiana* Dole) and Scots pine (*Pinus sylvestris* var. mongolica Litvinov). Many tree and understory species in the south-slope ecoregions are highly adapted to fire and readily resprout after crown fire. Soils are medium loam textured and well drained.

The north-slope ecoregions (ecoregions 17–21) account for 40% of forests in our study area and are dominated by dahurian larch, white birch, and Korean spruce (*Picea koriensis* Nakai). Vegetation regeneration after fire and harvest in the north-slope ecoregions was slower than that in the south-slope ecoregions (Cai et al. 2012). Soils are light clay

textured and moderately drained and have a thick layer of humus.

The ridge-top ecoregions (ecoregions 22–26) are located in the south and southwest part of our study area, representing 17% forests in the Huzhong Natural Reserve and the Huzhong Forestry Bureau. Dwarf pine (*Pinus pumila* (Pall.) Regel) is a dominant evergreen shrub in the ridge-top ecoregions. Soils are medium loam textured, acidic, and moderately drained, and the Podzolic horizon was 5–10 cm thick.

More than 40% of forestland in Huzhong Forestry Bureau has been harvested since the 1970s. Timber harvesting has significantly altered stand age and affects tree species composition. Surface fire is the dominant fire type across most of the study area, except for occasional crown fire in high elevation areas in the Huzhong Natural Reserve. Lightning accounts for 48% of ignitions in our study area, over 75% of which occur in summer. In the Huzhong Forestry Bureau and Huzhong Natural Reserve, human- and lightning-caused fires burned 352,889 ha forest from 1967 to 2005 (Liu et al. 2012).

Fire and harvest regulate forest structure and determine boreal forest succession trajectories. In the Great Xing'an Mountains, pioneer species (white birch and aspen) establish quickly after crown fire or high-intensity harvest. The shade-tolerant species (larch and Scots pine) then regenerate and coexist with pioneer species over the next 50–100 years. As forest succession continues, the short-lived pioneer species (120–150 years) gradually die and give way to longer-lived shade-tolerant species (250–300 years). Finally, boreal forests



Fig. 1 The geographic location of our study area and 26 ecoregions. Nonforested areas are not simulated in this study. HZNR: Huzhong Natural Reserve



will succeed to a climax community characteristic of coniferous larch forests in this region. Over the long term, fire and harvest resulted in forests dominated by dahurian larch, Scots pine, and white birch in the Great Xing'an Mountains. The stand age structure of these forests is affected primarily by fire and harvest. Mean stand age in our study area is 60 years old.

#### 2.2 The coupled modeling framework

In this study, we used LANDIS PRO to assess the effects of fire and harvest on stand structure and aboveground carbon because it can simulate forest succession, seed dispersal, fire, and harvest over long time periods (Wang et al. 2014a). The LANDIS PRO model tracks the number and density of trees by species age cohort for each ecoregion, which makes it possible to estimate aboveground carbon using diameter at breast height (DBH) and stand density. However, LANDIS PRO cannot simulate soil carbon and nitrogen cycling because it lacks a biogeochemistry module. The LINKAGES model incorporates tree physiological processes, demographic processes, microbial processes, and hydrologic processes to simulates soil organic carbon and nitrogen dynamics (Post and Pastor 1996). LINKAGES was selected for this study because it can simulate soil organic carbon stocks under fire and harvest by coupling with the LANDIS PRO model (Huang et al. 2017). Specifically, we apply LANDIS PRO to simulate the dynamics of stand structure under succession, fire, harvest, and fire-harvest scenarios. Then, we use outputs of LANDIS PRO (number of trees, age, and DBH by each ecoregion) as stand parameters in LINKAGES, which predicts soil organic carbon at the ecoregion level (Fig. 2).

#### 2.2.1 LANDIS PRO model and parameterization

LANDIS PRO is a forest landscape model that simulates forest succession under natural (e.g., fire, wind, insect, and disease outbreaks) and anthropogenic (e.g., harvest and fuel management) disturbances (Wang et al. 2016). The LANDIS PRO model tracks the number and density of trees by each species age cohort at the pixel level in addition to simulating growth, mortality, self-thinning, and species competition. The competition intensity is quantified as growing space occupied (GSO), which is estimated by Reineke's stand density index and pixel size (Wang et al. 2014a). These attributes can be used to calculate basal area or stocking percentage from tree volume and the number of trees. In LANDIS PRO, model parameters are compatible with commonly measured forest inventory and analysis data for easy parameterization, calibration, and validation (Wang et al. 2014b).

The fire module simulates fire occurrence, fire spread, and fire effects based on an ignition probability distribution, mean return interval, and mean size. Temporal patterns of fire regimes are simulated by a hierarchical fire frequency model,

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which divides fire occurrence into fire ignition and fire initiation (Yang et al. 2004). For each fire initiation, LANDIS PRO uses two percolation algorithms to estimate fire spread across the landscape from ignition points (Yang et al. 2008). The fire effect module determines which species' age cohorts are killed in a burned pixel (He and Mladenoff 1999). Fire effects are quantified as fire intensity that is determined by the quantity and quality of fuel in a pixel.

In the harvest module of LANDIS PRO, timber harvesting is simulated based on management area and stand maps. Management areas provide boundaries to constrain specific harvest events. Each management area is divided into stands that are the smallest contiguous units for harvesting operations (Fraser et al. 2013). In LANDIS PRO, trees harvested in a given stand are determined by stand age, basal area, stand stocking percent, and species composition.

Inputs for LANDIS PRO included raster maps (150 m resolution for this study) of ecoregion, species composition, species life history attributes, species establishment probability (SEP), and maximum growing space by each ecoregion. The species composition map was derived from China national forest inventory data and a forest type map that consists of the number of trees in each species age cohort. The forest type map was a GIS layer comprising boundaries of 46,862 stands that included forest composition information within each stand. China national forest inventory data, obtained from the Huzhong Forestry Bureau, contained numbers of trees by age class and DBH by species. The forest inventory plots were evenly distributed across the study area. In this study, we integrated the forest type map (polygon) and China national forest inventory data (point) to derive numbers of trees by age class for each species in the initial forest composition map. Species life history attributes (Table 1) were obtained from the Scientific Database of China Plant Species (http://db.kib.ac. cn) and previous studies (Li et al. 2013; Luo et al. 2015). Species establishment probability (SEP) was estimated from a previous study in this region (Luo et al. 2015).

Fire return interval, mean fire size, and fire ignition probability were estimated based on a database of historical fire records (1967–2005). In this study, the ecoregion map is used as fire regime input map to reflect the geographic differences in fire regimes. Fire return interval, mean fire size, standard deviation of fire size, and fire ignition probability of each ecoregion are provided in Table S2. According to the forestry management plan of the Huzhong Forest Bureau, our study area was divided into three management areas: forbidden harvest areas, restricted harvest areas, and commercial harvest areas (Chen et al. 2015). Forbidden harvest areas cover 56% of the forested areas in our study area and include the Huzhong National Natural Reserve and other conservation areas. In the restricted harvest areas and commercial harvest areas, larch, birch, and aspen are allowed to be harvested, whereas harvest of Scots pine, spruce, poplar, willow, and

Fig. 2 The framework for coupling LANDIS PRO and LINKAGES models to predict aboveground carbon and soil organic carbon



dwarf pine was not allowed. The main harvest method was stem-only clear-cutting (Table 2). The minimum stand age to harvest was 40 years. Three percent of the stands were harvested in the restricted harvest areas every 10 years (350 years rotation) and 5% were harvested in the commercial harvest areas every 10 years (200 years rotation). The minimum stand age to harvest was 40 years. When a harvest event was triggered within a management area, stands were chosen using a stocking level ranking algorithm (Fraser et al. 2013). Adjacent stands are not harvested for at least 10 years (Luo et al. 2014). In this study, we assumed that only legal harvesting was carried out, and the current fire and harvesting parameters did not vary during the whole simulation period (Table 2).

#### 2.2.2 LINKAGES V2.2 model and parameterization

LINKAGES v2.2 is a forest ecosystem model that is designed to simulate the dynamics of soil organic carbon and nitrogen evolved from LINKAGES1.0 (Wullschleger et al. 2003). LINKAGES v2.2 includes daily weather data, litter decomposition, mineralization, soil moisture, evaporation, tree species birth, growth, and mortality subroutines. These subroutines allow soil nutrient cycles to interact with species composition (Dijak et al. 2016). In LINKAGES v2.2, species composition affects the dynamics of soil organic carbon because species differ widely in how their litter affects soil nitrogen availability (Post and Pastor 1996). The main parameters of LINKAGES v2.2 include daily maximum and minimum temperature, daily precipitation, daily mean incoming solar radiation, daily mean wind speed, growing season degree-days, number of trees, diameter at breast height, age, number of soil layers, thickness of soil layer, soil layer field capacity, soil layer wilting point, soil organic matter, and soil nitrogen content.

Daily weather data for LINKAGES v2.2 were derived from the China national daily meteorological monitoring dataset (1975–2000; http://data.cma.gov.cn). We obtained soil parameters (thickness of soil layer, organic matter, nitrogen, rock fragment, clay and sand content) from a national soil database (http://www.soil.csdb.cn). Soil water-holding capacity and wilting point were estimated by rock fragment and clay and sand content (Table S3). The species parameters (Table 3) were derived from a previous study in northeastern China (He et al. 2005). Within each ecoregion, the number of trees, age, and DBH were obtained from the output of the LANDIS PRO model.



 Table 1
 Major species life history attributes for the LANDIS PRO model

Species	LONG <sup>a</sup>	MTR <sup>b</sup>	SHD <sup>c</sup>	FIRE <sup>d</sup>	EFFD <sup>e</sup>	MAXD <sup>f</sup>	VGP <sup>g</sup>	MAXDBH <sup>h</sup>	MAXSDI <sup>i</sup>	NPGS <sup>j</sup>	CARBONCO <sup>k</sup>	SEP <sup>1</sup>
Dahurian larch	300	20	2	5	50	400	0	70	600	10	0.4234	0.288
White birch	150	15	1	3	-1	2400	1	30	690	30	0.421	0.153
Korean spruce	300	30	4	1	100	300	0	60	520	10	0.48	0.06
Scots pine	250	25	2	4	100	300	0	70	560	10	0.412	0.175
Aspen	120	10	1	1	-1	2400	1	50	680	30	0.3921	0.01
Poplar	180	12	1	2	-1	-1	1	50	780	20	0.4144	0.013
Willow	250	12	2	1	-1	-1	0.9	50	780	20	0.43	0.018
Dwarf pine	250	15	3	3	90	100	0	20	340	5	0.51	0.225

<sup>a</sup> LONG, longevity of the species (years)

<sup>b</sup> MTR, maturity age of the species in year

<sup>c</sup> SHD, shade tolerance (1-5, 1 = least tolerant; 5 = most tolerant)

<sup>d</sup> FIRE, fire tolerance (1-5, 1 = least tolerant; 5 = most tolerant)

<sup>e</sup> EFFD, species effective seeding distance (m), -1 represents unlimited effective seeding range

<sup>f</sup>MAXD, species maximum seeding distance (m), -1 represents unlimited effective seeding range

<sup>g</sup> VGP, probability of vegetative propagation

<sup>h</sup> MAXDBH, maximum diameter at breast height of species in centimeters

<sup>i</sup> MAXSDI, maximum stand density (number of 25.5 cm trees per hectare)

<sup>j</sup>NPGS, number of potential germination seeds

k CARBONCO, carbon coefficient. This coefficient represents the amount of carbon in the wood of each species

<sup>1</sup>SEP, species establishment probability. The establishment of the potential sprouting seedlings is determined by this coefficient

#### 2.3 Simulation design and data analysis

To assess the long-term effects of fire and harvest on boreal forest carbon stocks, we designed four simulation scenarios: (1) succession-only scenario (fire and harvest were not simulated), (2) fire scenario (wherein fire and succession were simulated with the current fire regimes), (3) harvest scenario (whereby harvest and succession were simulated, with harvest regime reflecting current harvest practices in the Huzhong Forestry Bureau), and (4) fire-harvest scenario (whereby fire, harvest, and succession were simulated to reflect current conditions). The four simulation scenarios were simulated from year 2000 to 2150 at 10-year time steps with five replications to estimate uncertainties in the simulation results (Liang et al. 2012).

To quantify the effects of fire and harvest on carbon stocks, we compared the simulation results from the four simulation scenarios. In order to examine the temporal variation in fire and harvest effects, our simulation results were analyzed separately for the short term (0–20 years), medium term (30– 50 years), and long term (60–150 years). We used analysis of variance (ANOVA) to test the significance of the differentiation among fire and harvest and their combined effects over the short term, medium term, and long term. The Tukey's honest significant difference method was used for post hoc analyses (p = 0.05). To analyze the effects of fire and harvest on the spatial distribution of boreal forest carbon stocks, we compared the spatial changes of fire and harvest and their combined effects on carbon stocks at 50, 100, and 150 years. Three typical ecoregions (ecoregion-1, ecoregion-10, and ecoregion-25) were used to analyze the spatial variability of fire and harvest effects.

The "relaimpo" package (Gromping 2006) in R statistical software was used to calculate the relative importance of fire and harvest on boreal forest carbon stocks. This package

of	Species	Age	Cutting	Proportion area of ha	Deceneration	
		(years)	method	Restricted harvest areas	Commercial harvest areas	Regeneration
	Dahurian larch	120-300	Clear-cut	3	5	Natural
	White birch	60-150	Clear-cut	3	5	Natural
	Aspen	40-120	Clear-cut	3	5	Natural

The proportion of area harvested is a proportion of management area that will be harvested per decade

 
 Table 2
 The main parameters of the harvest module





#### Table 3 The major species parameters of the LINKAGES v2.2 model

Species	DMAX <sup>a</sup>	DMIN <sup>b</sup>	B3 <sup>c</sup>	B2 <sup>c</sup>	SWITCH <sup>d</sup>	D3 <sup>e</sup>	FROST <sup>f</sup>	FWT <sup>g</sup>	$TL^h$	RTST <sup>i</sup>	FRT <sup>j</sup>
Dahurian larch	1900	400	0.35	63.62	FFFFT	0.424	-35	248	12	1	1
White birch	3100	600	0.52	73.23	FFFFT	0.412	-40	248	4	0.8	1
Korean spruce	2500	600	0.38	68.07	FTFFF	0.11	-35	440	11	1	3
Scots pine	1900	700	0.37	59.08	TTTFF	0.6	- 30	440	12	1	3
Aspen	3000	800	0.66	78.77	TTFFT	0.333	-26	440	7	0.5	1
Poplar	1600	600	0.35	63.62	TTFFT	0.1	-35	440	7	0.5	1
Willow	2400	600	0.29	57.26	TTFFT	0.08	-35	248	7	0.5	1
Dwarf pine	1400	300	1.16	46.30	TTFFF	0.2	-40	248	12	1	3

<sup>a</sup>DMAX, degree day maximum for each species (degree)

<sup>b</sup> DMIN, degree day minimum for each species (degree)

<sup>c</sup> B2 and B3 are growth scaling parameters for each species

<sup>d</sup> SWITCH, reproduction switches

<sup>e</sup>D3, drought tolerance (fraction of growing season, %)

<sup>f</sup>FROST, minimum January temperature tolerated (degree)

<sup>g</sup> FWT, leaf weight per unit crown area  $(100 \text{ g/m}^2)$ 

<sup>h</sup> TL, leaf litter quality class (1–12 are the leaf litter; 13 is root litter; 14 and 15 are fresh wood from tress less than or greater than 10 cm; 16 is twig litter; 17 is well-decayed wood not contain humus)

<sup>i</sup> RTST, root-shoot ratio for each species

<sup>j</sup> FRT, foliage retention time (years)

implements the weighted averages with data-dependent weights method (pmvd) for assessing relative importance of predictor variables in multiple linear regressions. It also computes  $R^2_{pmvd}$ , or the average contribution of each variable to the overall  $R^2$  across all possible orderings, and thus provides a unique decomposition of the explained variance when predictors are correlated (Belmaker and Jetz 2011). The relative importance of fire and harvest was normalized (divided by  $R^2$ ) to sum to 1 (Carvalhais et al. 2014). With this method, we analyzed two response variables (aboveground carbon and soil organic carbon) at the short term, medium term, and long term. The R statistical software v3.2.2 (R Core Team 2015) was used in data analyses.

**Data availability** The data generated or analyzed during this study are included in this published article and its supplementary information files.

## **3 Results**

# 3.1 The effects and relative importance of fire and harvest and their combined effects on aboveground carbon

Aboveground carbon increased with simulation time at the landscape level under the succession-only scenario, fire, harvest, and fire-harvest scenarios (Fig. 3). Compared to succession-only scenario, aboveground carbon was reduced

by  $2.3 \pm 0.6$  t/ha in the fire scenario,  $6.0 \pm 1.4$  t/ha in the harvest scenario, and  $8.0 \pm 1.9$  t/ha in the fire-harvest scenario. The largest aboveground carbon reduction occurred in the fire-harvest scenario, accounting for 22% of the average value of the simulated aboveground carbon in the succession-only scenario. In the harvest scenario, aboveground carbon was 10 and 15% lower than that in the fire and succession-only scenarios, respectively. At the landscape level, the combined effects of fire and harvest on aboveground carbon were smaller than the sum of their effects in the fire scenario and the harvest scenario.



Fig. 3 Total landscape mean aboveground carbon for succession-only, fire, harvest, and fire-harvest scenarios over 150 years



The effects of fire and harvest on aboveground carbon varied among the three time periods (Table 4). In the short term (years 0–20), the effects of fire on aboveground carbon were significantly lower than those in the medium and long term (p < 0.05). The effects of fire and harvest on aboveground carbon did not vary significantly between the medium and long term (p > 0.05). At medium term, the effects of harvest on aboveground carbon were  $5.7 \pm 0.4$  t/ha, which were significantly higher than in short term. The combined effect of fire and harvest on aboveground carbon differed significantly among the three time periods (p < 0.05).

The relative importance of fire on aboveground carbon was 30.9, 26.8, and 14.2% in the three time periods, respectively (Fig. 4). The relative importance of harvest on aboveground carbon was greater than fire (69.1, 73.2, and 85.8% in the three time periods, respectively). The relative importance of harvest on aboveground carbon increased with simulation time; however, the relative importance of fire declined over 150 years.

# 3.2 The effects and relative importance of fire and harvest and their combined effects on soil organic carbon

At the landscape level, soil organic carbon increased with simulation year irrespective of fire and harvest (Fig. 5). Compared to the succession-only scenario, fire reduced soil organic carbon by  $5.4 \pm 3.1$  t/ha, harvest by  $5.8 \pm 3.3$  t/ha, and fire-harvest by  $6.5 \pm 3.4$  t/ha over 150 years. The largest soil organic carbon reduction occurred in the fire-harvest scenario, accounting for 8% of the simulated value in the succession-only scenario. In the fire and harvest scenarios, soil organic carbon was 6 and 7% lower than that in the succession-only scenario. The combined effects of fire and harvest on soil organic carbon were less than the sum of the separate effects of fire and harvest.

The effects of fire and harvest on aboveground carbon were significantly different at three time periods (Table 5). In the short term, the effects of fire on soil organic carbon were  $0.48 \pm 0.41$  t/ha, which were significant lower than those in the medium and long term (p < 0.05). Over the long term, the effects of harvest on soil organic carbon were  $7.7 \pm 2.3$  t/ha, which were significantly higher than those in the short term and medium term. The combined effect of fire and harvest on soil organic carbon differed significantly among the three time periods (p < 0.05).

The relative importance of harvest on soil organic carbon was greater than fire (Fig. 6). The relative importance of harvest on soil organic carbon was 86.4, 76.1, and 62.0% in the three time periods, respectively. Meanwhile, the relative importance of fire was 13.6, 23.9, and 37.0% in the short, medium, and long term. The relative importance of fire on soil carbon increased over time. In contrast, the relative importance of harvest on soil organic carbon decreased continuously over the 150-year simulations.

# 3.3 The spatiotemporal pattern of fire and harvest and their combined effects on boreal forest carbon stocks

In the three time periods, the spatial patterns of fire and harvest effects on boreal forest carbon stocks significantly differed among the fire, harvest, and fire-harvest scenarios (Fig. 7). After 150 years, the area affected by fire and harvest was 13% higher than it was after 50 years of simulation. The effects of fire on carbon stocks mostly occurred in the northwest (Huzhong Natural Reserve) and the ridge ecoregion over 800 m elevations. The main concentration of carbon stocks reduced by harvest occurred in commercial and restricted harvest areas, which included 45% of forest in our study area.

The effects of fire and harvest on boreal forest carbon stocks varied in different ecoregions (Fig. 8). In the terrace ecoregion (ecoregion-1), low fire frequency and relatively low harvest intensity led to less carbon stocks removed by these disturbances. In the south-slope ecoregion (ecoregion-10), relatively higher fire and harvest frequency tended to lower aboveground carbon and soil organic carbon over 150 years (Fig. 7). In the ridge top ecoregion (ecoregion-25), fire had greater effects on aboveground carbon and soil organic carbon than harvest, due to the frequency of fire being greater than the frequency of harvest in this area. These divergent responses were largely dependent on the intensity of fire and harvest among different ecoregions.

# 4 Discussion

In this study, we examined the effects of fire, harvest, and their spatial interactions on boreal forest carbon stocks of the Great Xing'an Mountains. Our simulation results showed that

Table 4Reductions inaboveground carbon after fire andharvest (unit: t/ha)

Period	Fire effects	Harvest effects	Combined effects
Short term (0–20 years)	$0.9 \pm 0.4 \; (3.0\%) \; a$	2.6 ± 1.1 (8.7%) a	3.4 ± 1.4 (10.5%) a
Medium term (30-50 years)	$2.2\pm 0.3~(6.5\%)~b$	$5.7 \pm 0.4 \ (16.4\%) \ b$	$7.3 \pm 0.5 \ (25.3\%) \ b$
Long term (60-150 years)	$2.5 \pm 0.4 \; (6.1\%) \; b$	$6.8 \pm 0.5 \ (16.7\%) \ b$	$9.1 \pm 0.2 \ (26.9\%) \ c$

Note: "a," "b," and "c" indicate significant differences (p < 0.05) between time periods, testing using analysis of variance (ANOVA) and Tukey's honest significant difference method





**Fig. 4** Relative importance of fire and harvest on aboveground carbon (*t*/ha) at three time periods (short term 0–20 years; medium term 30–50 years; long term 60–150 years)

recurrent fires, harvests, and their spatial interactions have strong effects on boreal forest carbon stocks in northeastern China. At the landscape level, the long-term effects of fire and harvest on carbon stocks were greater than the short-term effects. The extent of carbon loss by fire and harvest depended on the intensity of these disturbances. Harvest plays a more important role in affecting boreal forest carbon stocks than fire, which may be attributed to the high intensity of clear-cuts. The spatial distribution of boreal forest carbon stocks was affected by the spatial pattern of fire and harvest in different ecoregions.

Our simulation results showed that fire, harvest, and their spatial interactions decreased aboveground carbon and soil organic carbon over the short term. This result is generally consistent with previous studies (Kashian et al. 2006; Thornley and Cannell 2004) in boreal forests. However, the short-term effects of fire and harvest on boreal forest carbon stocks in our study are smaller than those in some field studies. For example, Hu et al. (2016) found that fire reduced aboveground carbon by 9.5–16.5 t/ha in the larix gmelinii-grass



Fig. 5 Total landscape mean soil organic carbon for succession, fire, harvest, and fire-harvest scenarios over 150 years

stands of the Great Xing'an Mountains during 1987–2014. The discrepancy in the extent of fire effects between our study and those of Hu et al. (2016) was likely due to the differences in estimation methods. In their study, the effects of fire and harvest on aboveground carbon were calculated as the difference in carbon between disturbed and undisturbed plots. Beyond the plot level, disturbances rarely entirely burn or harvest an entire landscape due to environmental heterogeneity, and thus, plot-level assessments tended to overestimate disturbance effects (Dean et al. 2004). This discrepancy suggests that accounting for disturbance intensity and extent as well as landscape configuration is important for regional-scale carbon stock assessments. Meanwhile, some field studies indicated that additional inputs of litter associated with fire and stem-only harvesting cause a short-term increase in soil carbon stocks (Johnson and Curtis 2001). However, most field studies were conducted within a relatively short period of time (e.g., < 20 years) and, therefore, may not capture the long-term impacts of disturbances on soil organic carbon (Clarke et al. 2015). Since our simulation time step is 10 years, most such short-term spikes in postdisturbance soil organic matter were smoothed out.

Our results showed that the effects of fire, harvest, and their spatial interaction on boreal forest carbon stocks increased from the short to long term. Aboveground carbon was reduced by harvest  $6.8 \pm 0.5$  t/ha over the long term, which was about three times greater than the short-term effects. Our findings contradict some previous studies suggesting that forest recovery after fire and harvest can offset the effects of these disturbances on forest carbon stocks over the long term (Chen and Shrestha 2012; Nalder and Wein 1999). This is because fire and harvest usually occur in different stands at different times that alter the distribution of stand ages across a landscape. More frequent fire and harvest will promote a higher proportion of young stands in the landscape. Young stands in the landscape tend to store less carbon than old and mature stands, but have higher carbon sequestration rates (Alexandrov 2007; Kashian et al. 2006). In addition, many boreal forests experience repeated disturbances associated with relatively short fire return intervals and harvest rotations (Bergeron et al. 2001; Bose et al. 2013). Over the long term, repeated fire and harvest not only increase carbon losses from fire emissions and timber removals (Cheng et al. 2013) but also alter the carbon recovery trajectory of boreal forest ecosystems (Johnstone and Chapin 2006; Steenberg et al. 2012). Our findings reinforce the importance of long-term studies when assessing the effects of fire and harvest on boreal forest carbon stocks and demonstrate the utility of ecological simulation models for assessing long-term disturbance impacts.

The results from our study showed that harvest was more important than fire in affecting boreal forest carbon stocks over 150 years (Figs. 4 and 6). There are several reasons for such results. First, the percentage of area affected by timber



Table 5Reductions in soilorganic carbon after fire andharvest (unit: t/ha)

Period	Fire effects	Harvest effects	Combined effects
Short term (0–20 years)	$0.4 \pm 0.2 \; (0.3\%) \; a$	$0.5 \pm 0.2 \; (0.4\%) \; a$	0.6 ± 0.3 (0.5%) a
Medium term (30-50 years)	$1.8 \pm 0.6 \ (1.7\%) \ b$	$2.0 \pm 0.9 \; (1.9\%) \; b$	$2.4\pm 0.9~(2.2\%)~b$
Long term (60–150 years)	$7.2 \pm 2.3 \ (6.5\%) \ c$	$7.7 \pm 2.3 \ (7.0\%) \ c$	$8.5 \pm 2.3 \ (7.7\%) \ c$

Note: "a," "b," and "c" indicate significant differences (p < 0.05) between time periods, testing using analysis of variance (ANOVA) and Tukey's honest significant difference method

harvesting is much larger than that of fire (Fig. 7). Second, clear-cutting not only removes more aboveground biomass than fire but also markedly increases soil temperature, and this might lead to increased litter decomposition and hence increased carbon release from mineral soil by heterotrophic respiration (Zha et al. 2009). Third, the rate of tree regeneration after clear-cutting is slower than after fire (Cai et al. 2013; Xu et al. 2006), which may delay forest carbon recovery over the long term (Carlson et al. 2012). This implies that harvest is the primary disturbance controlling carbon stocks in boreal forests of the Great Xing'an Mountains.

Our simulation demonstrated that the effects of fire and harvest are less than the sum of carbon loss from the fire and harvest-only scenarios. This is because timber harvesting reduces fuel loads, thereby decreasing the risk of high-severity fire (Bury 2004). The forest gaps caused by high-intensity harvest have negative impacts on fire spread and tree damage (Kiefer et al. 2016). Our results suggest that harvest may come at the expense of less carbon on the landscape, but reduces the amount of carbon loss to high-severity fires. This implies that moderate timber harvesting may be a useful tool for limiting carbon loss from catastrophic fires in the fire-prone forests.

Meanwhile, our results showed that the effects of fire and harvest on carbon stocks varied among ecoregions due to differences in fire and harvest regimes. This is similar to a previous study that showed the divergent response of carbon dynamics among ecoregions and climate change scenarios was due to the spatial pattern of soils, tree species, and



**Fig. 6** Relative importance of fire and harvest on soil organic carbon at three time periods (short term 0–20 years; medium term 30–50 years; long term 60–150 years)

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land-use histories (Scheller et al. 2012). In the terrace ecoregion (ecoregion-1), fire and harvest had smaller effects on boreal forest carbon stocks than they did in other ecoregions. The terrace ecoregion is located in a plain which is close to the Huma River, roads, and settlements, where fires were mostly of human origin and most fires were small in size due to suppression (Wu et al. 2015). In addition, most forests in the terrace ecoregion belonged to areas where timber harvesting was forbidden and consequently had limited effects, whereas fire had greater effects on carbon stocks than harvest in the ridge top ecoregion (ecoregion-25). Most of the forest in the ridge top ecoregion was over 800 m elevation. Fires in this ecoregion were primarily of lightning origin. Most of the fires were not suppressed because of the remoteness, and the fire sizes were large (Wu et al. 2014). The effects of harvest were smaller than fire in this ecoregion due to approximately 70% forest in the ridge-top ecoregion being in forbidden harvest areas. These results suggested that the spatial distribution of boreal forest carbon stocks is influenced by the spatial pattern of fire and harvest among different ecoregions.

Boreal forest carbon stocks are the results of the interactions between climate, stand age, species composition, soil nutrients, and disturbances. It is challenging to accurately predict boreal forest carbon stocks under different disturbance regimes over long time periods. Ecological models provide a unique scientific approach for long-term prediction of the effects of fire and harvest on forest carbon stocks. The simulation results of ecological models can illustrate general trends in forest carbon stocks and provide insight into how fire and harvest affect forest dynamics in the future (He et al. 2011, 2017; Shifley et al. 2006). However, validation of simulation results requires long-term spatial and temporal field data, which are not readily available. In this study, forest inventory data and mature forests in the Huzhong Natural Reserve were used to validate model performance. The results of the model validation showed that simulated values were comparable to observed values derived from the forest inventory data and Huzhong Natural Reserve (Huang et al. 2017).

Our study quantified the long-term effects of fire, harvest, and their spatial interactions on boreal forest carbon stocks based on a series of reasonable assumptions. First, this study is based on the assumption that fire regimes in our study area remained unchanged over the simulation period. Over the long term, the frequency and intensity of fire is closely related



to climate variables, that may impact forest carbon dynamics (Running 2008). As the frequency and intensity of fire is expected to increase in the future due to climate change (Flannigan et al. 2005), fire will further reduce the carbon stocks in boreal forests of the Great Xing'an Mountains.

Second, we assumed that forest management planning and timber harvest methods remained unchanged. The harvest practices were strictly carried out according to forest management planning in the LANDIS PRO model. However, timber harvesting without permission was not simulated in our study



Fig. 8 Effects of fire and harvest on aboveground carbon and soil organic carbon in three example ecoregions. The top panels show aboveground carbon removed by fire or harvest. The top panels show aboveground

carbon removed by fire or harvest. The middle panels show soil organic carbon that was removed by fire or harvest. The bottom panels show the percentage of ecoregion area that burned or was harvested



but is an observed activity. Our study may underestimate the effects of harvest on boreal forest carbon stocks because we have not considered illegal logging activities. Besides, we did not track the carbon stocks of forest products, so we might overestimate the effects of harvest. Third, species life history attributes and the relationships between species establishment probability and environmental factors (topography, elevation, and climate) were constant, which in reality likely vary across different environmental conditions (Levesque et al. 2013). These factors may alter the effects of fire and harvest and their spatial interactions on boreal forest carbon stocks. Fourth, we did not include carbon stocks of forest understory in our simulations. This was because the understory accounts for less than 3% of total aboveground carbon in our study area. A previous field study showed that fire has no significant effects on shrub and herb carbon stocks in the boreal forests of the Great Xing'an Mountains (Yang 2015).

The initial forest conditions in 2000, from a legacy of decades of timber harvest, fire, and forest succession, could have a great effect on subsequent simulation results. Two major sources of uncertainties of forest inventory data included (1) decade-long time span (1990-2000) of the inventory data and (2) misregistration of inventory data to forest stands. Our underlining hypothesis is that the 9-year time difference in forest age and stand-level information mismatch does not affect the long-term simulation results. First, our simulations were conducted at a 10-year time step in which for example, a 40-year-old stand and a 49-year-old stand were treated as 40-year-old stand. Since trees are long lived with average longevity ranging from 150 to 300 years, the 9-year age difference in the inventory data should not affect our overall results. Second, Xu et al. (2004) found that the effects of cell-level uncertainties had little effect on long-term LANDIS model simulation results. Therefore, we believe that stand-level information mismatch in our study should not alter our overall results and conclusion.

The effects of fire and harvest on soil physical and chemical processes were not taken into consideration in this study. Previous studies showed that root decomposition after harvesting and charcoal sequestration after fire play an important role in soil organic carbon dynamics (Johnson et al. 1991; Wardle et al. 2008). In boreal forests, fire converts a fraction of the burning vegetation to charcoal carbon. Charcoal carbon is slow to decompose, and the formation of this reserve therefore creates a long-term soil carbon sink (Czimczik et al. 2005). However, the rate of charcoal sequestration and root decomposition after disturbances is not well understood (Fahey and Arthur 1994; Ohlson et al. 2009). Thus, the effects of root decomposition and charcoal sequestration on soil carbon stocks were not taken into consideration in this study. Although these assumptions and limitations created uncertainly in our simulation results, the results from our study still can be used to identify general trends in the response of boreal

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forest carbon stocks to fire, harvest, and their spatial interactions under current climate conditions, disturbance regimes, and forest management practices.

# **5** Conclusions

In this study, we found that boreal forest carbon stocks are significantly altered by fire, harvest, and their spatial interactions. Our simulation results suggested that the long-term effects of fire and harvest on boreal forest carbon stocks are greater than those short-term effects at the landscape level. The relative importance of fire and harvest on boreal forest carbon stocks was dependent on the intensity of these disturbances. Because timber harvesting reduces the amount of carbon loss to fires, the combined effects of fire and harvest on boreal forests are less than the sum of the separate effects of fire and harvest. Spatial variation in the effects of fire and harvest on boreal forest carbon stocks is caused by diverse disturbance regimes among different ecoregions. These results have important implications for forest managers to design and implement sustainable forest fire and harvest strategies to maintain carbon stocks of boreal forests.

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### **Compliance with ethical standards**

**Conflict of interest** The authors declare that they have no conflict of interest.

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