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\*CORRESPONDENCE Himanshu Bargali ⊠ himanshubargali44@gmail.com Aseesh Pandey ⊠ draseeshpandey@gmail.com

RECEIVED 07 December 2023 ACCEPTED 12 April 2024 PUBLISHED 03 June 2024

#### CITATION

Bargali H, Pandey A, Bhatt D and Sundriyal RC (2024) Loss of carbon stock in the forests of Uttarakhand due to unprecedented seasonal forest fires. *Front. For. Glob. Change* 7:1352265. doi: 10.3389/ffgc.2024.1352265

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# Loss of carbon stock in the forests of Uttarakhand due to unprecedented seasonal forest fires

Himanshu Bargali<sup>1,2</sup>\*, Aseesh Pandey<sup>1</sup>\*, Dinesh Bhatt<sup>2</sup> and R. C. Sundriyal<sup>3</sup>

<sup>1</sup>Center for Biodiversity Conservation and Management, GB Pant National Institute of Himalayan Environment, Almora, Uttarakhand, India, <sup>2</sup>Department of Zoology and Environmental Science, Gurukula Kangri (Deemed to be University), Haridwar, Uttarakhand, India, <sup>3</sup>Department of Forestry and Natural Resources, Hemvati Nandan Bahuguna Garhwal University, Srinagar, Garhwal, India

Unprecedented seasonal forest fires pose a significant threat to the carbon stocks of diverse ecosystems, particularly in regions like Uttarakhand, west Himalaya. Understanding the impact of varying fire frequencies on different forest types is crucial for effective conservation and management strategies. This study aims to assess the loss of carbon stock in three distinct forest types—Sal, Pine, and Mixed across an elevation gradient in Uttarakhand, facing unprecedented seasonal forest fires. By investigating pre- and post-fire conditions, analyzing biomass dynamics, and mapping fire frequencies, the research aims to provide insights into the complex interplay of fire regimes and forest resilience. The investigation covers vegetation analysis, biomass assessment, and fire frequency mapping. Biomass and carbon stock calculations were carried out using a non-destructive sampling method. Fire frequency maps were generated using Landsat satellite imagery spanning a decade, integrating MODIS hotspot data for classification. The study reveals distinct patterns in biomass changes across Sal, Pine, and Mixed forests in response to varying fire frequencies. Sal forests exhibit resilience to low-intensity fires, while Pine forests show higher sensitivity. Carbon stock contributions of dominant species varied significantly, with Sal and Chir-Pine forests emerging as crucial contributors. High fire frequencies lead to substantial carbon stock reduction in all forest types. The findings emphasize the sensitivity of aboveground biomass to fire frequency, with significant carbon stock loss observed in higher fire frequency classes. The study underscores the importance of nuanced conservation strategies tailored to distinct forest types and species characteristics. This research provides valuable insights for policymakers, forest managers, and conservationists in formulating targeted conservation and management approaches.

#### KEYWORDS

forest fire, carbon loss, carbon stock, Himalayan forest, Uttarakhand, Indian Himalayan region

# Introduction

Forests provide a wide array of essential goods and services that sustain life (Bisht et al., 2018), encompassing fuel, timber, food (Joshi and Negi, 2011), bioproducts (Shmulsky and Jones, 2019), greenhouse gas regulation (Ribeiro-Kumara et al., 2020), air quality (Baró et al., 2014), water supply (Cassiano and Ferraz, 2022), carbon storage (Fahey et al., 2010;

Rai et al., 2018), nutrient cycling (Sharma and Sharma, 2004), and genetic and species diversity (Connell and Orias, 1964; de Groot et al., 2002; Singh, 2002). The preservation of forests has become a prominent policy objective for environmental safeguarding (Koskela and Karppinen, 2021), and sustenance of their critical ecological and climatic contributions (Humpenoder et al., 2014; Capellesso et al., 2021; Maren and Sharma, 2021). Uttarakhand Himalaya, known for its rich floral and faunal diversity (Anthwal et al., 2010), geology, and ecology, also considered a repository of biological (Bargali et al., 2022a) and cultural diversity (Negi, 2010), warranting conservation and sustainable management for future generations (Kumar et al., 2015; Mathela et al., 2020).

However, according to the India State of Forest Report (FSI, 2021), the forest cover in Uttarakhand is only 45.44% (Kumar and Hole, 2021; Bargali et al., 2022a), significantly below then the recommended forest cover (66%) for hill states in India (National Forest Policy, 1952). The prevalence of forest fires in the most hill states is one of the contributing factors to forest degradation (Babu et al., 2016; Ranjan, 2018; Babu, 2019; Bargali et al., 2022). Fire, a natural and recurrent phenomenon (Vasconcelos et al., 2017), exerts both positive and negative effects on forests (Certini, 2005). Positive outcomes involve enhancing biodiversity and stimulating plant growth (Bhatt et al., 2019; He et al., 2019), while negative impacts encompass habitat destruction, alterations in soil chemistry, and reduction in forest biomass (Bargali et al., 2017; Bowd et al., 2021). Grasping the consequences of fires on forests is vital for their conservation and management (Naveh, 1994; Bowman and Balch, 2013), especially in regions where fires occur frequently and with intensity (Bargali et al., 2022a). In Uttarakhand Himalaya, fires are traditionally perceived as disruptive and harmful (Rawat et al., 2017), with their role in local forest ecosystems historically marginalized (Gururani, 1996; Niklasson et al., 2010; Bargali et al., 2020).

Studies of paleoecology and fire history indicate that human presence in the landscape has increased the frequency of fire events in the past (Bahuguna and Upadhay, 2002; Whitlock et al., 2010; Negi and Dhyani, 2012; Abrams and Nowacki, 2019). Fire is recognized as one of the Earth's most powerful agents of ecological change (Allendorf and Hard, 2009; Stenseth and Dunlop, 2009), with far-reaching ecological implications (Harrison et al., 2010). Forest fires entail significant losses (Hansen et al., 2010), encompassing human lives (Pausas and Keeley, 2009) economic consequences (De Mendonça et al., 2004), depletion of forest resources (Sadowska et al., 2021), destruction of biogeocenosis (Zhukov, 1976), and severe impacts on soil (Verma and Jayakumar, 2015) leading to loss and erosion post-fire (Scott et al., 2009; Jhariya et al., 2012). Additionally, forest fires contribute to greenhouse gas emissions, changes in climate patterns (Kurniawan et al., 2022), and the loss of ecosystem values (Lecina-Diaz et al., 2021) and environmental services (Lee et al., 2015). Key impacts include reduced floral productivity (Flower and Gonzalez-Meler, 2015) changes in regeneration rates (Deb and Sundriyal, 2008; Sathya and Jayakumar, 2017; Boucher et al., 2020), and the loss of various endemic and endangered fauna species (Laurance and Useche, 2009; Yule, 2010; Sati and Bandooni, 2018; Ward et al., 2020).

In Uttarakhand Himalaya, forest fire intensities are anticipated to increase due to rising temperatures and global warming (Ahmad and Goparaju, 2018). Most forest fires in this region are of anthropogenic origin (Bhandari et al., 2012), stemming from both accidental and intentional sources (Semwal and Mehta, 1996; Chauhan et al., 2018;

Fulé et al., 2021). Fire behavior, which involves the ignition of fuels (Burgan, 1984), flame development (Vahabi et al., 2015), and fire spread (Countryman, 1964), is influenced by the interplay of fuels (Bessie and Johnson, 1995) weather, and topography in forest fires (Agee, 1996). Additionally, the type of forest, calorific value of fuels, dead plant matter, and litter plays a significant role (Hakkila and Parikka, 2002; Cornelissen et al., 2017).

In recent decades, forest fires have become more frequent and widespread in the Himalayan state of Uttarakhand (Singh et al., 2016; Bargali et al., 2017; Ahmad and Goparaju, 2018; Fulé et al., 2021). These forests have historically experienced fires for thousands of years (Gadgil, 1992; Anthwal et al., 2010; Bargali et al., 2024), with fire being a significant influence on landscape patterns and species diversity (Delcourt and Delcourt, 1997; Bhandari et al., 2012; Bargali et al., 2022b). The prevalence of plant populations offers insight into species' capacity to respond to disturbances, including forest fires (Rodríguez-Trejo, 2014; Miller et al., 2019). In Uttarakhand, the dominance of Chir-pine (Pinus roxburghii) covers approximately 28% of the forest area (Fulé et al., 2021), contributing to the acceleration of forest fires (Kumar et al., 2019; Ray et al., 2019). The altitudinal zone of Chir pine forests, between 1,000 and 1800 meters above sea level, is considered a fire-prone area in the study region (Negi, 2019). Anthropogenic activities, such as the collection of non-timber forest products and agricultural practices, have also played a role in forest fires in the study area (Bargali et al., 2020; Bhandari and Bijlwan, 2020; Saha et al., 2023).

Forests act as the key carbon pool (C) and store more carbon per unit area compared to other terrestrial ecosystems (Zhao et al., 2014). Forest cover is around 30% of the Earth's total surface area, and contains 19% of the Earth's overall biomass and carbon pool (Kindermann et al., 2008; FAO and UNEP, 2020); forests having greater carbon in biomass and soil than in any natural ecosystem and atmosphere (Pan et al., 2011; Zhao et al., 2014; Jiang et al., 2022). Tropical and temperate forests are global centers of biodiversity, which play an important role in the regulation of the global and regional C cycles (Poorter et al., 2015). Among various forests, tropical forests fixes higher biomass, and serve as a major potential sink of carbon due to high species diversity and high net primary production (Malhi et al., 1999; Kothandaraman et al., 2020). Approximately 767 million hectares (25%) area covered by temperate forest globally of the total land surface, and they storage about 14% of total carbon (Pan et al., 2011). Currently, climate change is a global concern, and forests plays vital role in climate change regulation and mitigation through reducing CO<sub>2</sub> concentrations in the atmosphere (Streck and Scholz, 2006; Scholz and Huber, 2017; Bracki, 2019; Ali et al., 2020; Burman et al., 2021). Thus, estimation of carbon stocks of different forest ecosystems would help in appropriate decision-making on carbon (C) management. This kind of information also contributes toward atmospheric carbon reduction targets as part of international obligations (UNFCCC, 2014; Sahu et al., 2016; Mayer et al., 2020). The amount of total biomass stored in a forest indicates the quantity of C that can be sequestered to meet the emission targets (Brown et al., 1999; Raha et al., 2020).

Carbon sequestration is among the most important ecosystem services provided by forest ecosystems that play an important role in global climate mitigation. For example, in a recent study Tolangay and Moktan (2020) reported that the Indian Himalayan region (IHR) sequesters about 65 million tonnes of carbon each year. This has leaded to conclude that 'the Himalayan forests have the potential to mitigate climate change and global warming'. We hypothesize that seasonal forest fires in Uttarakhand Himalaya lead to significant losses in carbon stocks within forest ecosystems. This hypothesis is based on the understanding that forest fires contribute to habitat degradation, alter vegetation dynamics, and release stored carbon into the atmosphere, thereby impacting the overall carbon sequestration capacity of the forests. With the above background and hypothesis in mind, a first-of-its-kind study was conducted in the Himalayan region on forest fires, it was aimed to analyze the loss of carbon stock in the forest of Uttarakhand due to unprecedented seasonal forest fires.

# Materials and methods

## Study area

The investigation encompassed three primary forest types along the elevation gradient (300–1700 m a.s.l) in Uttarakhand state of Western Himalaya, India. Among the chosen forest stands, the Motichur, Haridwar site (Site-I) was characterized by the dominance of *Shorea robusta* Gaertn. (Sal), while the Mona, Nainital range (Site-II) was dominated by *Pinus roxburghii* Roxb. Chir-Pine and the Dhaulchina, Almora site (Site-III) was primarily dominated by *Quercus leucotrichophora A. Camus* (Banj-oak). Geologically, all the sites (stands) were situated in the lesser Himalayan region (Valdiya, 1980). All the selected forests are managed by the forest department. A comprehensive overview of the geographical and ecological characteristics of the study sites are provided in Table 1.

## Vegetation analysis

Vegetation analysis was conducted by deploying random quadrats within identified forest stands both before and after a fire event. A total of 30 ( $10 \times 10$  m) quadrats were randomly placed for the investigation. The Circumference at Breast Height (CBH) of each tree species was measured at 1.37 m above the ground. The collected field data underwent analysis for various phytosociological parameters, including density, frequency, basal area, and species richness (Mishra, 1968). Standard phytosociological methods, as outlined by Ellenberg

and Muller-Dombois (1974), were employed for obtaining field data. Microsoft Excel 2019 was used to analyze phytosociological data.

## **Biomass analysis**

Biomass and productivity were assessed using a non-harvesting sampling method, with 1.0 ha areas sampled in each site. All trees with a diameter less than 30 cm in the sampled quadrat were recorded and measured for Circumference at Breast Height (CBH) at 1.37 m above the ground. Yellow paint was used to mark all trees for long-term monitoring. Carbon stock was considered as approximately half (0.475) of biomass, and carbon sequestration was estimated as half of productivity, following the approach by Magnussen and Reed (2004). Trees measured in permanent plots during the first year (2017) were re-measured in the subsequent years (2018 and 2019).

Biomass for different components (bole, branch, twig, foliage, stump root, and fine roots) in the first year (Y1), second year (Y2), and third year (Y3) was calculated using the regression equation of the allometric method (Rawat and Singh, 1988; Rana et al., 1989; Basuki et al., 2009; Navar, 2009; Dabi et al., 2021): Y=a+b·lnX. Here, ln denotes the natural log, Y represents the dry weight of the component (kg), X is CBH (cm), a is the y-intercept, and b is the slope of the regression (Table 2). Net productivity was determined by the difference in biomass between 2 years:  $\Delta Y = Y2 - Y1$ . Where  $\Delta Y$  is the net primary productivity, Y1 is the biomass in the first year, and Y2 is the biomass in the second year. Biomass and carbon stock were computed in Megagrams per hectare (Mgha-1). Carbon stock and sequestration were assumed to be half of the total estimated biomass of each tree species, following the methodology by Magnussen and Reed (2004):  $stock = Biomass \cdot 0.475$ ;  $sequestration = NPP \cdot 0.475$ stock = Biomass-0.475; C sequestration = NPP-0.475.

## Fire frequency mapping

Forest fire frequency maps were created for the state of Uttarakhand by using Landsat 5, 7, and 8 satellite imagery spanning a decade (2011–2020). The MODIS hotspot data

TABLE 1 General features of surveyed forests in Uttarakhand, Western Himalayan region.

S. no.	Parameter (s)	Sal forest	Chir-Pine forest	Mixed forest
1.	Forest location	Motichur, Haridwar	Mauna, Nainital	Dhaulchina, Almora
2.	Altitude (m a.s.l)	320	1,109	1,642
3.	Latitude	29°29′ 29.912″	29°23′ 42.447″	29°27′ 56.309″
4.	Longitude	79°32′ 21.770″	79°39′ 09.842″	79°39′ 07.832″
5.	Canopy species	Shorea robusta	Pinus roxburghii	Quercus leucotrichophora
6.	Soil moisture (%)	$18.04 \pm 3.91$	$22.15 \pm 4.09$	$28.90 \pm 5.43$
7.	WHC (%)	$49.27 \pm 2.23$	$54.54 \pm 3.22$	$62.78 \pm 4.32$
8.	рН	$6.44\pm0.13$	$6.43\pm0.34$	$5.89 \pm 0.76$
9.	SOC (%)	$1.22 \pm 0.034$	$2.32\pm0.12$	3.42±0.23
10.	N (%)	$0.12 \pm 0.012$	$0.21 \pm 0.091$	$0.41\pm0.023$
11.	Soil texture (%)	Loamy sand	Sandy clay	Sandy loam

WHC, water holding capacity; N, nitrogen; SOC, soil organic carbon.

TABLE 2 Allometric relationships between the biomass of tree components (Y,  $kg^{-1}$  tree) and circumference of tree at breast height (X, cm at 1.37 m height) for different tree used.

Biomass (Kg tree <sup>-1</sup> )	Intercept (a)	Slope (b)	r <sup>2</sup>			
Pinus roxburghii						
Bole	-6.418	2.598	0.985			
First order brance	-9.833	2.978	0.979			
Other branches	-9.338	2.630	0.963			
Foliage	-6.111	1.872	0.952			
Stamp root	-7.220	2.448	0.978			
Lateral root	-9.161	2.593	0.974			
Fine root	-9.102	2.069	0.938			
Interspecies or mixed forest						
Bole	-0.861	1.425	0.915			
Branch	-0.908	1.327	0.907			
Twig	-0.506	1.028	0.796			
Leaf	-1.106	1.042	0.755			
Stump root	-0.098	0.948	0.789			
Lateral root	-2.346	0.997	0.724			
Fine root	-2.874	0.529	0.722			
Sal forest						
Boie	-2.832	1.976	0.980			
Branch	-2.037	1.501	0.922			
Twig	-2.688	1.463	0.980			
Leaf	-1.736	1.175	0.960			
Total	-1.789	1.892	0.980			

The equation is: Ln Y = a + by Ln X; where Ln is natural log, a Intercept of Y and b slope or regression coefficient (based on Tewari et at., 1985; Rana et al., 1989). Interspecies equation is significant at p < 0.05, rest are significant at p < 0.01.

(MOD14) from NASA's Fire Information for Resource Management System (FIRMS) official website were downloaded and integrated with shape-file data representing fire points (Bargali et al., 2023). The satellite images for specific sites were extracted, geo-corrected, and classified into four frequency classes—namely, low fire, moderate fire, and high fire frequency classes (Figure 1). Supervised classification was applied to categorize the extracted and geo-corrected satellite images of chosen study sites, including Sal Forest, Chir-Pine Forest, and Mixed Forest. Using the data management tool in ArcMap software, a fishnet was generated, and fire points and raster data were spatially joined to produce the fire frequency map. A total of 16,206 fire points were recorded and classified into four fire frequency classes for the entire state. To exclude fire points in settlements, the settlement layer was overlaid onto the fire point layer, ensuring that only fire points within forest regions were considered for analysis. The fire frequency was further divided into four classes-low fire, moderate fire, and high fire frequency classes, as illustrated in Figure 1. To maintain objectivity, a 1 hectare area for each forest type was selected within each fire frequency class.

## Data analysis

The statistical analysis involved paired *t*-tests independently for both aboveground and belowground carbon stock components. Using Statistical Package for the Social Sciences (IBM, SPSS) Statistics 26 package.

## **Results**

The paired *t*-test analysis was conducted on the pre-fire and post-fire carbon stock of three forest types: Sal Forest, Chir-Pine Forest, and Mixed Forest. For the aboveground carbon stock, the mean differences and their respective standard errors (SE or  $\pm$ ) were calculated. In Sal Forest, the mean difference was 12 Mg ha<sup>-1</sup> with an SE ( $\pm$ ) of 8 Mg ha<sup>-1</sup>, resulting in a t-statistic of 1.29 [t(2) = 1.29, p = 0.29], indicating a lack of statistical significance. However, in Chir-Pine Forest, the mean difference was 20 Mg ha<sup>-1</sup> with an SE (±) of  $10 \text{ Mg ha}^{-1}$ , yielding a t-statistic of 2.00 [t(2) = 2.00, p = 0.12], which was statistically significant. Conversely, Mixed Forest showed a mean difference of 8 Mg ha<sup>-1</sup> with an SE  $(\pm)$  of  $4 \, \text{Mg} \, \text{ha}^{-1}$ , also not reaching statistical significance [t(2) = 2.00, p = 0.12]. Moving to Belowground carbon stock, Sal Forest had a mean difference of  $4 \text{ Mg ha}^{-1}$  (SE = 2) and a *t*-statistic of 2.00 [t(2) = 2.00, p = 0.12], indicating no significant difference. Chir-Pine Forest exhibited a mean difference of  $6 \text{ Mg ha}^{-1}$  (SE = ± 4) with a t-statistic of 1.50 [t(2) = 1.50, p = 0.23], also not statistically significant. However, Mixed Forest showed a notable mean difference of  $7 \text{ Mg ha}^{-1}$  (SE ± 2) with a significant t-statistic of 3.50 [t(2) = 3.50, p = 0.04], indicating a significant change in Belowground carbon stock post-fire.

# Impact of pre-and-post fire frequencies on different forest types

## Sal forest

Pre-fire, tree biomass in the Sal forest ranged from 270 Mg ha<sup>-1</sup> (Low fire) to 219 Mg ha<sup>-1</sup> (High fire) (Figure 2). Among all fire classes, aboveground biomass was highest in the low fire class (270 Mg ha<sup>-1</sup>), followed by moderate fire (250 Mg ha<sup>-1</sup>), and high fire (230 Mg ha<sup>-1</sup>). Post-fire, the tree biomass in the Sal forest exhibited changes, ranging from 265 Mg ha<sup>-1</sup> (Low fire) to 219 Mg ha<sup>-1</sup> (High fire) (Figure 2). Aboveground biomass remained highest in the low fire class (265 Mg ha<sup>-1</sup>), followed by moderate fire (243 Mg ha<sup>-1</sup>), and high fire (219 Mg ha<sup>-1</sup>).

## **Chir-Pine forest**

Pre-fire, tree biomass in the Chir-Pine forest ranged from  $200 \text{ Mg} ha^{-1}$  (Low fire) to  $122 \text{ Mg} ha^{-1}$  (High fire) (Figure 2). Among all fire classes, aboveground biomass was highest in the low fire class  $(200 \text{ Mg} ha^{-1})$ , followed by moderate fire ( $160 \text{ Mg} ha^{-1}$ ), and high fire ( $130 \text{ Mg} ha^{-1}$ ). Post-fire, the tree biomass in the Chir-Pine Forest displayed changes, ranging from  $198 \text{ Mg} ha^{-1}$  (Low fire) to  $123 \text{ Mg} ha^{-1}$  (High fire) (Figure 2). Aboveground biomass remained highest in the low fire class ( $198 \text{ Mg} ha^{-1}$ ), followed by moderate fire ( $155 \text{ Mg} ha^{-1}$ ), and high fire ( $123 \text{ Mg} ha^{-1}$ ).



#### Mixed forest

Pre-fire, tree biomass in the Mixed forest ranged from 170 Mg ha<sup>-1</sup> (Low fire) to 123 Mg ha<sup>-1</sup> (High fire) (Figure 2). Among all fire classes, aboveground biomass was highest in the low fire class (170 Mg ha<sup>-1</sup>), followed by moderate fire (150 Mg ha<sup>-1</sup>), and high fire (135 Mg ha<sup>-1</sup>). Post-fire, the tree biomass in the Mixed Forest showed changes, ranging from 164 Mg ha<sup>-1</sup> (Low fire) to 123 Mg ha<sup>-1</sup> (High fire) (Figure 2). Aboveground biomass remained highest in the low fire

class (164 Mg ha  $^{-1}$  ), followed by moderate fire (142 Mg ha  $^{-1}$  ), and high fire (123 Mg ha  $^{-1}$  ).

#### Sal forest

The analysis of carbon stock changes in the Sal forest, categorized by different fire frequency classes, revealed distinct patterns in the aftermath of fires. In the low fire frequency class, a relatively minor reduction of 5 MgC in carbon stock was observed. This suggests that low-intensity fires





have a limited impact on carbon storage within the Sal forest. Moving to the moderate fire frequency class, a more substantial decrease of 7 MgC in carbon stock was noted. This indicates that as the fire intensity increases, the magnitude of the impact on carbon stocks also increases. Notably, in the high fire frequency class, the most significant reduction was observed, with a noteworthy decline of 11 MgC in carbon stock. This underscores the heightened vulnerability of the Sal forest to carbon loss in the presence of frequent and intense fires (Figure 3).

#### Chir-Pine forest

The examination of carbon stock dynamics within the Chir-Pine forest, classified by varying fire frequency classes, revealed distinctive trends in the aftermath of fires. In the low fire frequency class, a minimal reduction of 2 MgC in carbon stock implies that low-intensity fires have a limited impact on carbon storage within the Chir-Pine forest. Transitioning to the moderate fire frequency class, a more pronounced decrease of 5 MgC in carbon stock suggests an escalating impact as fire intensity increases. Notably, in the high fire frequency class, the most substantial reduction was observed, with an 8 MgC decline in carbon stock (Figure 3).

### Mixed forest

Examining the carbon stock dynamics in the mixed forest across different fire frequency classes reveals distinctive patterns prior to and after fire incidents. In the low fire frequency class, there was a marginal reduction of 6 MgC in carbon stock, indicating a relatively modest impact of low-intensity fires on carbon storage within the mixed forest. Moving to the moderate fire frequency class, a more significant decrease of 8 MgC in carbon stock was observed, suggesting an escalating impact as fire intensity increases. Notably, in the high fire frequency class, the most substantial reduction was evident, with a noteworthy decline of 12 MgC in carbon stock (Figure 3).

## Species wise contribution of carbon stock

#### Sal forest

The carbon stock contributions of different tree species in the Sal forest are presented in the Figure 4. In the Low fire, *Shorea robusta* exhibited the highest carbon stock contribution among the species with a value of 36 MgC ha<sup>-1</sup>, followed by *Mallotus philippensis* with 16 MgC ha<sup>-1</sup>. *Syzygium cumini, Butea monosperma*, and *Aegle marmelos* contributed 18, 12, and 14 MgC ha<sup>-1</sup>, respectively. Under Moderate fire conditions, *S. robusta* maintained its leading position, contributing 33 MgC ha<sup>-1</sup>, followed by *M. philippensis* with 13 MgC ha<sup>-1</sup>. *S. cumini, B. monosperma*, and *A. marmelos* contributed 14, 9, and 7 MgC ha<sup>-1</sup>, respectively. In the High fire class, the contribution of *S. robusta* declined to 12 MgC ha<sup>-1</sup>, while *M. philippensis* became the highest contributor with 14 MgC ha<sup>-1</sup>. *S. cumini, B. monosperma*, and *A. marmelos* contributed 16, 16, and 6 MgC ha<sup>-1</sup>, respectively.

#### **Chir-Pine Forest**

In the Chir-Pine forest, the carbon stock contributions of two dominant tree species, *Pinus roxburghii* and *Myrica esculenta*, are outlined in the Figure 4. Under Low fire conditions, *P. roxburghii* dominated the carbon stock contribution with a value of 73.68 MgC ha<sup>-1</sup>, while *M. esculenta* contributed 26.31 MgC ha<sup>-1</sup>. In the Moderate fire scenario, *P. roxburghii* maintained a substantial carbon stock contribution of 46.18 MgC ha<sup>-1</sup>, surpassing *M. esculenta*, which contributed 28.12 MgC ha<sup>-1</sup>. In the High fire class, both species experienced a reduction in carbon stock contributions. *P. roxburghii* contributed 16.12 MgC ha<sup>-1</sup>, while *M. esculenta* contributed 14.13 MgC ha<sup>-1</sup>.



### Mixed forest

In the mixed forest, the carbon stock contributions of three dominant tree species—*Quercus leucotrichophora, Myrica esculenta, and Pinus roxburghii*—are detailed in the Figure 4. Under Low fire conditions, *Q. leucotrichophora* emerged as the primary contributor to carbon stock with 38.35 MgC ha<sup>-1</sup>. *M. esculenta* and *P. roxburghii* contributed 13.69 and 18.26 MgC ha<sup>-1</sup>. *M. esculenta* and *P. roxburghii* contribution 13.69 and 18.26 MgC ha<sup>-1</sup>. *M. esculenta* followed closely with 18.1 MgC ha<sup>-1</sup>, while *Q. leucotrichophora*'s contribution decreased to 16.46 MgC ha<sup>-1</sup>. In the High fire scenario, *M. esculenta* became the top contributor and yielded a carbon stock of 26.26 MgC ha<sup>-1</sup>. *Q. leucotrichophora* and *P. roxburghii* contributed 8.16 and 4.26 MgC ha<sup>-1</sup>, respectively.

## Discussion

Aboveground biomass (AGB) stands as a crucial quantitative parameter for assessing the health and sustainability of forest ecosystems (Joshi et al., 2022). AGB reflects the cumulative biomass of trees and vegetation above the soil surface and serves as a key indicator of the ecosystem's carbon storage capacity. Our investigation into AGB changes in three diverse forest types—Sal, Chir-Pine, and Mixed—in response to varying fire frequencies provides valuable insights into the complex interplay of factors influencing forest dynamics. The observed alterations in tree biomass were influenced by a multitude of factors inherent to forest ecosystems. These include stand age, species composition, forest type, size class of trees, site conditions, rainfall patterns, edaphic factors, and elevation (Peichl and Arain, 2006; Gairola et al., 2011). Each of these factors contributed uniquely to the overall dynamics of AGB, shaping the structure and function of the forest. Our results affirmed that fire, whether occurring as a low frequency event or high frequency, exerts a discernible influence on AGB. The observed changes in AGB following fires in the Sal, Chir-Pine, and Mixed forests underscore the sensitivity of this parameter to variations in fire frequency. The impact of fire on AGB was particularly pronounced in higher fire frequency classes, as demonstrated by the substantial reductions in carbon stock. The Sal forest, known for its resilience to lower intensity fires, exhibited a gradient of AGB loss corresponding to increasing fire frequency. The Chir-Pine forest, characterized by higher sensitivity to fire, displayed more substantial biomass reductions across all fire frequency classes. In the Mixed forest, the diversity of tree species contributed to varied responses to fire and further emphasized the complexity of factors influencing AGB dynamics. The paired t-test analysis also revealed significant changes in the carbon stock dynamics post-fire, with Chir-Pine Forest showing a notable shift in Aboveground carbon stock and Mixed Forest exhibiting a significant change in Belowground carbon stock. The comparison of carbon stock contributions across the three distinct forest types-Sal, Pine, and Mixed-revealed interesting patterns and species-specific dynamics in response to varying fire intensities. Examining the data collectively provides valuable insights into the overall carbon sequestration capacity and highlights the contributions of individual tree species in each forest type. In the Sal forest, S. robusta consistently emerged as a significant contributor to carbon stock, leading the pack across all fire frequencies. Its remarkable resilience in maintaining high carbon stock levels showcased its importance in carbon sequestration, underscoring the need for its preservation in forest management strategies. Contrastingly, the Chir-Pine forest displayed a different pattern, with P. roxburghii playing a pivotal role in carbon stock contributions. This species demonstrated a robust ability to sequester carbon, particularly evident in low and moderate fire scenarios. While contributing substantially, M. esculenta does not match the carbon stock levels of P. roxburghii. mixed forest presented a unique scenario, where The Q. leucotrichophora, M. esculenta, and P. roxburghii showcased varying contributions across different fire intensities. Q. leucotrichophora dominated in low fire frequency, whereas P. roxburghii took the lead in moderate fires, and M. esculenta became the primary contributor in high fire frequency. It was evident that the species response to fire intensity played a crucial role in determining their carbon sequestration capacity. Across all forests, P. roxburghii consistently exhibited significant carbon stock contributions, showcasing its resilience in the face of different fire regimes. This emphasized the importance of understanding the unique characteristics of each species for effective forest management and conservation.

## Conclusion

Our findings emphasize the need for a comprehensive understanding of the interactions between fire, forest type, and above ground biomass (AGB) dynamics. The identification of specific fire frequency classes associated with minimum and maximum biomass losses provides critical information for targeted conservation strategies. Acknowledging the multifaceted nature of AGB changes which contributes to the development of nuanced and effective forest management approaches, ensuring the resilience and sustainability of these diverse ecosystems in the face of evolving fire regimes. The findings suggest that preserving and promoting certain key/local/ native species, such as Shorea robusta in Sal forests and Pinus roxburghii in Chir-Pine forests, can contribute significantly to maintaining or enhancing carbon rather than new species. These insights will be instrumental for policymakers, forest managers, and conservationists in devising strategies to ensure the sustainability and resilience of diverse forest ecosystems in the context of changing fire dynamics and global climate patterns.

## Data availability statement

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

## Author contributions

HB: Conceptualization, Data curation, Investigation, Methodology, Software, Writing – original draft, Writing – review &

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editing. AP: Supervision, Writing – original draft, Writing – review & editing. DB: Supervision, Writing – review & editing. RS: Writing – review & editing.

## Funding

The author(s) declare that no financial support was received for the research, authorship, and/or publication of this article.

## Acknowledgments

The authors extend their appreciation to the Director of G. B. Pant National Institute of Himalayan Environment (NIHE) for institutional support. We also recognize the Department of Zoology and Environment Science at Gurukula Kangri (Deemed to be University), Haridwar, for their logistical assistance. Special thanks are given to Rohit Kumar, IIT, Roorkee for his valuable suggestions in the fire frequency map.

## **Conflict of interest**

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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