Effects of vegetation and topography on biomass patterns post-pinewood nematode disturbance in forest landscapes of southern Jiangxi Province, China

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Abstract

**Background:** Assessing changes in forest ecosystems, especially forest biomass changes that occur due to disturbances, is essential for improving global carbon estimates.

*Bursaphelenchus xylophilus* is an important pest that harms *Pinus thunbergii* forests, causing a large number of pine trees to wither and resulting in great losses in *Pinus thunbergii* forest biomass. Studying the factors affecting *P. thunbergii*) biomass losses caused by *Bursaphelenchus xylophilus* is of great significance in research on forest ecosystem health and climate change. In this study, taking Nankang District, Ganzhou City, Jiangxi Province as the research area, based on field survey data combined with a random forest model, the relative importance of influencing factors on the biomass loss of *P. thunbergii* due to the *Bursaphelenchus xylophilus*-caused disease was analyzed.

**Results:** The results revealed the following conclusions: (1) topographic and slope conditions significantly affected the *P. thunbergii*) biomass losses caused by the *Bursaphelenchus xylophilus*-related disease; (2) the two studied stand factors, vegetation type and vegetation coverage, had little effect on the *P. thunbergii*) biomass losses caused by *Bursaphelenchus xylophilus*; and (3) the marginal effect diagram showed that the elevation and slope were obviously related to biomass loss; biomass loss was positively related to elevation and negatively related to slope.

**Conclusions:** Our study demonstrated that topographical factors dominantly affect the spread of the *Bursaphelenchus xylophilus*-caused disease, in turn causing large *P. thunbergii* forest biomass losses. Therefore, topographic factors affect the prevention and control of the disease caused by *Bursaphelenchus xylophilus*.

1. **Background**

As an important part of terrestrial ecosystems, forest ecosystems play an irreplaceable role in maintaining biodiversity and regulating the global carbon balance (Kramer et al 1981; Trumbore et al 2006; He et al 2012). Forest biomass is an important indicator used to measure the function of forest ecosystems (Waring and Schlesinger 1985; Logan 2003; Villanueva et al. 2011). Assessments of forest biomass changes have always been an important component of research on global changes (Dixon et al. 1994; Forkuor G et al. 2020).

Insect pests and diseases constitute an important factor causing the loss of forest biomass (Keeling et al 2013). The World Food and Agriculture Organization (FAO) of the United Nations reported that approximately 40 million hectares of forests were affected by pests and diseases in 2015 (FAO 2020). Among these pests, the pine wood nematode [*Bursaphelenchus xylophilus* (Steiner & Buhrer) Nickle] causes a very dangerous forest disease (Jones et al. 2013) that has the characteristics of a high fatality rate, a wide range of damages, fast transmission and spread, being difficult to control, and so on. As a result, this species has been listed as the key object of forest pest quarantine measures in many countries and regions (e.g., Japan, South Korea, and Europe) (Kiyohara and Tokushige 1971, Kim et al. 2020). Pine
wood nematode disease has spread to 18 provinces since it was first discovered in China in 1982 (Yang et al. 2014). Although the Chinese government has launched active initiatives for the prevention and control of the pine wood nematode, the forest biomass losses caused by pine wood nematodes have remained significant for decades. For example, according to the 2020 report of the state forestry administration of the People's Republic of China (http://www.gov.cn/index.htm), the accumulation of wood loss caused by pine wood nematodes in China was approximately 0.68 million acres.

Scholars have studied the spatiotemporal variations (Ye and Giblin-Davis et al. 2007; Hu et al. 2011; Joana M. et al. 2012), pathogenic mechanisms (Mamiya and Kiyohara 1972; Mamiya 1983; Kikuchi et al. 2011), and transmission modes of pine wood nematodes (Linit et al 1988). For example, Katsumi Togashi et al. (2006) studied a transmission model of pine wood nematodes and concluded that the emergence of pine wood nematodes was closely related to the host longicorn species, artificial transportation, and climatic conditions. Research by Chinese scholars has mainly focused on the physiological factors and vectors of pine wood nematode disease (Zhang et al. 2007; Zhao et al. 2007, Huang et al. 2010; Zhao et al. 2016). For example, Long Pan et al. (2020) studied the population characteristics of pine wood nematodes under low-temperature environmental stress, and the results shown an enhanced adaptability of pine wood nematodes to low-temperature conditions (Pan et al. 2020). Pine wood nematodes are biological invasive species; their harm to forest ecosystems is greatly affected by environmental factors. The impacts of environmental factors on the ecological consequences (such as forest biomass losses) of pine wood nematodes need to be studied further.

The *Pinus massoniana* pine forests in the Nankang District, southern Jiangxi Province, are seriously affected by pine wood nematodes. In 2020, Nankang District was listed by the State Forestry and Grass Administration of China (http://www.gov.cn/) as a key epidemic area of pine wood nematode infection. After the pine wood nematodes invaded the trees, the needles first lost water, faded to green, and then browned; then, the whole plants withered, and all the needles turned red and yellow and, finally, died. The local government strictly followed the national “Technical Plan for Prevention and Control of Pine Wood Nematode” ([2018] No. 110) and implemented projects involving the cleanup and disposal of infected trees to ensure that no diseased trees or pine branches with diameters of 1 cm or more were left behind. Hidden hazards such as stumps were also removed (https://www.ganzhou.gov.cn/), resulting in the loss of forest biomass. The existing research on pine wood nematodes affecting *P. massoniana* in Nankang District has mainly focused on surveillance and control (https://www.ganzhou.gov.cn/). Research on the biomass loss caused by pine wood nematodes in this region has not yet been reported.

The amount of biomass lost after a pine wood nematode disturbance differs due to the vegetation and topographic conditions. The specific objectives of this study were as follows: (1) to analyze the differences in the degree of *P. massoniana* biomass loss caused by pine wood nematodes under different vegetation and topographic conditions; (2) to analyze the relative influences of vegetation and topographic factors on Masson pine biomass loss; and (3) to identify the key factors affecting the biomass loss caused by pine wood nematodes. The results are expected to provide managers with an understanding of the impact and mechanism of pine wood nematodes affecting regional forest biomass.
2 Methods

2.1 Study area

This study was conducted in Nankang District, Jiangxi Province, southern China (Fig. 1). The terrain in this region is dominated by mountains and hills and ranges in elevation from 106–1042 m. The climate is a subtropical humid monsoon climate with an average annual rainfall of 1443.2 mm; the monthly average temperature is 8.8°C in December and 28.6°C in July. The land area of Nankang District is 1722 km², of which 60.8% is covered by forests. The main tree species in the study area include Masson pine (*Pinus massoniana* Lamb.), fir (*Cunninghamia lanceolata* (Lamb.) Hook.), Schima superba Gardn. et Champ. and maple (*Staphylea forrestii* Balf. F.).

2.2 Overall study approaches

Through field sample surveys, the height and diameter at breast height (DBH) of each tree in each sample plot (20 m × 20 m in size) were obtained. Information on the terrain factors (slope, slope position, aspect, and elevation) and stand factors (canopy closure and forest type) of each plot was also obtained. Based on stand biomass models, the biomass of each tree in each sample plot was calculated. The Masson pine biomass loss was obtained using the proportion of the total biomass of the diseased wood. Duncan’s multiple comparison method was used to evaluate the differences in the biomass losses that occurred under different vegetation and terrain conditions. We used the random forest model to analyze the relative importance (to obtain the contribution rate of each factor to the biomass loss) and then identified the key factors. A marginal effect analysis was used to derive the trend of the influence of each factor on Masson pine biomass loss.

2.3 Sampling and experimental design

We obtained field data from 120 plots (20 m × 20 m in size) in September 2020 (Fig. 3). These plots were representative of the major forest types and topographic conditions in the study area (Table 1). The latitude and longitude at the center of each sample plot were recorded, and GPS equipment was used to record the slope, elevation, and aspect of each sample plot. The height and DBH of each tree in each 20 m × 20 m sample plot were investigated. The Masson pine forests in Nankang District were all aerially seeded from 1965 to 1975, so the forest stands and stand ages in the area are essentially the same. The two statistics had no practical significance, so the stand age and stand composition were not considered.
### Table 1
Plot information

| Index            | Slope | Elevation | Canopy closure | Aspect | Slope position | Forest type |
|------------------|-------|-----------|----------------|--------|----------------|-------------|
| Sequential statistics |      |           |                |        |                |             |
| Minimum          | 11    | 122       | 0.4            | -      | -              | -           |
| 1st quartile     | 16    | 163.8     | 0.7            | -      | -              | -           |
| Median           | 18    | 179       | 0.8            | -      | -              | -           |
| 3rd quartile     | 23    | 219.2     | 0.8            | -      | -              | -           |
| Maximum          | 45    | 474       | 0.9            | -      | -              | -           |
| Arithmetic mean  | 20.27 | 203.48    | 0.76           | -      | -              | -           |
| Standard deviation | 6.54 | 69.44     | 0.08           | -      | -              | -           |
| Skewness coefficient | 1.28 | 1.85      | -0.85          | -      | -              | -           |
| Kurtosis coefficient | 1.59 | 3.12      | 3.19           | -      | -              | -           |

### 2.4 Data preprocessing

In accordance with the “Technical Regulations for the Continuous Inventory of National Forest Resources (2014)”, we quantified the terrain and forest stand factors as follows (Table 2).

### Table 2
Data quantification

| Factor type   | Factor      | Classification                                      |
|---------------|-------------|-----------------------------------------------------|
| Terrain factors | Slope       | 1 gentle slopes (6–15°), 2 moderate slopes (16–25°), 3 steep slopes (26–35°), 4 very steep slopes (36–45°), and 5 dangerous slopes (≥ 46°) |
| Terrain factors | Aspect      | 1 sunny slope, 2. semi-sunny slopes, 3 semi-shady slopes, and 4 shady slopes |
| Terrain factors | Slope position | 1 uphill, 2 mid-slope, 3 downhill |
| Stand factors | Elevation   | 1 0–200 meters, 2 200–400 meters, 3 400–600 meters, 4 > 999 meters |
| Stand factors | Forest type | 1 pure Masson pine forest, 2 mixed forests |
| Stand factors | Canopy closure | 1 medium (0.4 ~ 0.69), 2 dense (≥ 0.7) |
(1) Terrain factors

The sample plot slopes were divided into five categories: 1 represented a gentle slope (6–15°), 2 represented a moderate slope (16–25°), 3 represented a steep slope (26–35°), 4 represented a very steep slope (36–45°), and 5 represented a dangerous slope (≥ 46°). In regard to the slope aspect, 1 was used to represent a sunny slope, 2 was used to represent a semi-sunny slope, 3 was used to represent a semi-shady slope, and 4 was used to represent a shady slope. The slope positions were divided into three categories: 1 for uphill, 2 for mid-slope, and 3 for a downhill position. The elevation was divided into four levels: 1 for elevations from 0–200 m, 2 for elevations form 200–400 m, 3 for elevations from 400–600 m, and 4 for elevations > 999 m.

(2) Stand factors

The forest types were classified into two categories: pure Masson pine forest and mixed Masson pine forest, where 1 represented a pure Masson pine forest and 2 represented a mixed forest with Masson pine. The canopy density was divided into greater than 0.7 and less than 0.7.

2.5 Calculation of forest biomass loss

A binary (tree height and DBH) standing tree biomass model (Table 3) from the “People's Republic of China Forestry Industry Standards” (LY/T2263-2014, LY/T2264-2014, LY/T2260-2016) was used to calculate the biomass of each tree in each survey plot (including those infected with and uninfected with pine wood nematode disease). The biomasses of the whole sample plots and the biomass of each pine tree infected with pine wood nematode disease were counted (Regulations on Continuous Inventory of National Forest Resources et al. 2014).

| Species                 | Binary model                                      |
|-------------------------|---------------------------------------------------|
| *Pinus thunbergii*      | $M_A = 0.066615D^{2.09317}H^{0.49763}$ (D ≥ 5)  |
|                         | $M_A = 0.117268D^{1.74179}H^{0.49763}$ (D < 5)  |
| *Cunninghamia lanceolata* | $M_A = 0.032718D^{2.11093}H^{0.60212}$ (D ≥ 5)  |
|                         | $M_A = 0.173824D^{1.07322}H^{0.60212}$ (D < 5)  |
| *Schima superba*        | $M_A = 0.12045D^{2.06446}H^{0.38265}$ (D ≥ 5)  |

**Note:** D: diameter at breast height; H: height

| Biomass of plagued trees/biomass of all trees in the sample plot x 100% |
2.6 Statistical analysis

All statistical analyses were conducted in R version 4.0.4. First, we used Duncan's multiple comparison method to analyze whether there was a significant difference in the biomass losses of Masson pine under different vegetation and topographic conditions. Duncan's multiple comparison method is a nonparametric comparison method that analyzes differences among groups and has been widely used in biological and ecological applications for multiple comparison analysis research (Warton et al. 2012, Kopelman et al. 2015). Duncan's multiple comparison method was performed using the laercio package in R language (Silva et al. 2010).

We used the random forest (RF) model to evaluate the relative effects of topographic and vegetation factors on *P. massoniana* biomass loss. The random forest is an integrated algorithm that uses the bootstrap resampling method to extract multiple samples from an original sample and generates an independent decision tree for each sample (Liaw et al. 2002). The random forest calculates the influence of each variable on the heterogeneity of the observed value at each node of the classification tree through the Gini index to compare the importance of each variable. The larger the value is, the greater the importance of the variable is. The randomForest package in R language was also used in this study (Breiman and Cutler et al. 2008).

A marginal effect analysis was used to reveal the trend of the contribution of each influencing factor to Masson pine biomass loss. In this study, the partialPlot function was used to plot the contribution of each influencing factor to the overall biomass loss.

3 Results

3.1 Statistical description

In all the surveyed sample plots, the biomass loss rate caused by pine wood nematodes was between 5% and 78%, with an average value of 19.3%. A standard deviation of 0.134 was recorded. The distribution was skewed to the right (skewness + 2.7), showing more positive values than negative values. There were 81 plots with biomass loss rates between 5% and 20% and 32 plots with biomass loss rates ranging from 20–34%. The loss rates of the remaining plots ranged from 34–78% (Table 4).
Table 4  
**Biomass loss information**

| Statistic              | Biomass loss rate |
|------------------------|-------------------|
| Minimum                | 0.057             |
| 1st quartile           | 0.11              |
| Median                 | 0.15              |
| 3rd quartile           | 0.22              |
| Maximum                | 0.77              |
| Arithmetic mean        | 0.19              |
| Standard deviation     | 0.13              |
| Skewness coefficient   | 2.69              |
| Kurtosis coefficient   | 7.87              |

**3.2 Differences in biomass loss rates among stands and topographic factors**

Duncan's multiple comparison analysis showed that significant differences existed in biomass losses among sample plots with different topographic and forest stand factors (Fig. 4). Significant differences were determined in biomass loss between steep slopes and very steep slopes; the biomass loss was higher on steep slopes. Sunny slopes and shaded slopes showed a statistically significant difference, and the biomass loss was higher on sunny slopes than on shaded slopes. Between mid-slope and upslope plots, the biomass loss was higher in the mid-slope position. Duncan's multiple comparison showed a significant difference in biomass losses between the elevations of 400–600 m and 0–400 m; the biomass loss was higher in the high hills than in those closer to sea level. Conversely, there was no obvious difference observed in biomass losses among different forest types or different canopy closures.

**3.3 The relative importance of different influencing factors**

A random forest model was used to calculate the relative importance of the influencing factors (Fig. 5). From the ranking of the impact factors shown in Fig. 5, we can see that the importance different among different impact factors. The factors influencing Masson pine biomass loss, in order of importance, were the elevation, slope, canopy closure, slope aspect, slope position, and forest type of the plots. The overall relative importance of the topographic factors were higher than those of the forest type factors.

**3.4 The marginal effects of each impact factor**

The marginal effect of each influencing factor on Masson pine biomass loss is shown in Fig. 6. As the elevation increased, the marginal effect curve also showed a rising trend, indicating a positive correlation between elevation and biomass loss; with an increase in the slope, the marginal effect curve of biomass loss showed a declining trend, indicating that the slope and biomass loss were negatively correlated. In
terms of the slope position, the marginal effect value of the middle slope position was the highest, while those of the downslope position and the upslope position were lower. The marginal effect first increased and then dropped with the changing slope position. Regarding the slope aspect, the highest marginal effect value was obtained for the sunny slope, and the differences among the values of the semi-sunny slope, semi-shady slope, and shady slope were very small. There was no obvious difference between the marginal effect values of the mixed forest and pure forest, and this factor had little impact on biomass loss. The changes in the marginal effect value between moderate canopy closure and dense canopy closure were similar to the forest type results; the difference between the two was very small impact of this factor on biomass volume loss was also small.

4 Discussion

As a statistical method, the random forest method is widely used in ecological research to address the complex relationships between multidimensional data and variables (Prasad et al. 2006; Morera et al. 2020). For example, Iverson L R et al. (2008) used a random forest model to analyze the potential habitats of 134 eastern tree species in the United States under six climatic conditions. Random forest classification and regression can process more complex ecological data than generalized linear models (De’ath et al. 2000). Other studies have also suggested that errors can gradually converge when the number of classification trees in a random forest decision tree becomes sufficiently large to improve the accuracy of the model (Breiman 2001). In studies of biological invasive species, researchers have confirmed that the most important variables influencing invasive species as derived by random forest methods were consistent with those reported in previous studies by scholars (Cutler et al. 2007).

This study showed that the average biomass loss rate of all sample plots was 19.3%, indicating that pine wood nematodes have a great impact on the regional forest ecosystem. Similar findings have also shown that biological invasions affect the productivity of forest ecosystems, and productivity also changes with biomass losses (Lovet et al. 2006). The biomass loss showed a right-biased direction, indicating that the overall biomass loss of the sample plots was relatively high. It is necessary to understand the formation mechanism of the regional environmental conditions that cause extreme biomass losses. In addition, many studies have also shown that Masson pine biomass displays a downward trend after a Masson pine ecosystem is affected by pine wood nematodes, and biomass loss continues to expand with artificial felling and replacement with other tree species (Spiegel and Leege 2013).

This study showed that the slope and elevation of the sample plots were the most significant factors influencing biomass loss (Fig. 5). With an increasing slope, the biomass loss caused by pine wood nematodes first increased and then decreased, and the overall trend increased. This was because an increase in slope reduces the accumulation of organic matter in soils, affecting the growth of the pest host, Pinus massoniana. In addition, an increase in slope usually reduces the soil water content and affects the occurrence of nematodes. For example, Suzuki and Kiyohara (1978) and other scholars (1983) obtained similar results after studying the emergence of pine wood nematodes under soil water stress. The elevation of the sample plots was positively correlated with the biomass loss caused by pine
wood nematodes; biomass loss did not occur at low elevations, the losses increased at mid-high elevations, and pests did not occur when the elevation continued to increase. The occurrences of wood nematodes yielded similar results. Elevation was positively correlated with the biomass loss caused by pine wood nematodes. This could be due to the transmission ability of the pests, the behavioral characteristics of the pine wood nematode carrier *A. longiflora*, and the temperature drop that occurs with an increase in elevation (Zhang et al. 2007). A growing number of studies have shown that temperature has a controlling effect on the occurrence of pine wood nematodes (Hashimoto et al. 1976). It was observed during the survey conducted in this study that the aspect and position of the slope of the sample plots also had certain impacts on the biomass loss. The biomass loss caused by pine wood nematodes differed among different aspects. The biomass loss was higher in the sunny aspect than that in the shaded aspect. It was found that pine wood nematodes harmed Masson (Dwinell et al. 2003); as a result, different slope positions had different effects on the biomass losses caused by pine wood nematodes. The biomass loss of the middle slope position was higher than those of the downslope position and the upslope position. The organic matter in the soils of the middle slope position accumulated to a greater extent than those upslope or downslope, and the temperature of the middle slope position was suitable for the pest insects, leading to high biomass loss rates occurring after pest infestations (Ryss et al. 2011). On the other hand, studies on pine wood nematodes niches in China and the established pine wood nematode spread model have also shown that features such as elevation and temperature are important factors affecting the spatial distribution of pine wood nematodes (Wu and Li 2009).

Figure 7 shows that the sample plots containing pine wood nematodes at elevations of approximately 250 meters accounted for a large proportion of the total number of sample plots, and the proportions of biomass losses were highest in these sample plots. No pine wood nematode pests were observed in the areas below 100 m or above 500 m, resulting in slight biomass losses.

The stand factor had little effect on the biomass loss caused by pine wood nematodes. Regarding the forest type and canopy density, whether the sample plot contained a pure forest or a mixed forest, the degree of pine wood nematode occurrence was essentially the same for both moderate-canopy-density and high-canopy-density plots. This result was similarly to the results reported in other studies. For example, pine wood nematodes have been observed finding and invading Masson pine trees in mixed forests within short distances (Yoshimura, Akiko et al. 1999, Togashi and Shigesada 2006). This is mainly due to the biological characteristics of pine wood nematode and the behavioral characteristics of the host longhorn beetle; because of these characteristics, pine wood nematodes can pass through short distances in non-Masson pine forests. When the density of a Masson pine forest is too low or too high and the host longhorn beetle density is too low, the emergence of pine wood nematodes becomes difficult (Kishi Y 1995; Takasu et al. 2000). The local government is supposed to treat infected Masson pine stands every year to reduce the spread of the Masson pine disease. However, it is difficult to construct operations in areas with high elevations, steep slopes, and high canopy closures. If infected areas are not controlled in the coming years, the spread of pine wood nematode disease will be aggravated, resulting in large losses.
This study showed that terrain factors are more important than stand factors in general; temperature and soil moisture are mainly affected by terrain factors, and temperature and soil moisture are the most important factors affecting the occurrence of pine wood nematodes. Mamiya and other researchers in Japan and Europe also arrived at similar conclusions (Rutherford, Mamiya et al. 1990; Ikegami and Jenkins 2018). During the investigation of 120 sample plots, it was found that topographical factors were more important than stand factors for pest control, and this research was significance in guiding management practices. In the actual prevention, control, and management of pine wood nematode disease, attention should be given to the healthy cultivation of Masson pine trees, and healthy pine trees should be maintained. The growth of pine trees requires water to improve the resistance of the trees to disease. A previous study also found that high biomass losses occurred approximately 100 meters around roads; this result may be related to the local furniture manufacturing industry and to other manual transportation practices in the study area.

In this study, the estimation of biomass loss from the *P. massoniana* disease caused by pine wood nematodes and the correlation analysis of biomass loss with environmental factors on the plot scale provided a reference for estimating large-scale pest-related biomass losses and spatiotemporal changes in pest-related biomass losses. However, climatic conditions also lead to the aggravation and spread of pine wood nematodes and biomass losses (Ena and Mamiya 1972; Mcgawley E C et al. 1985; Iki T et al. 2020). This paper did not analyze the correlations between meteorological conditions or climatic factors and the degree of pine biomass loss; therefore, the results obtained in this paper had certain limitations. In future work, we can obtain meteorological data to analyze biomass losses in Masson pine forests and analyze the Masson pine biomass losses caused by pine wood nematodes more comprehensively.

**5 Conclusions**

Pine wood nematodes pose a threat to forest ecosystems and forest biomass. To prevent the spread of pine wood nematodes, their transmission mechanism, including internal transmission factors and external environmental factors, must be determined. The determination of the transmission mechanism would not only help in understanding pine wood nematodes but would also provide information relevant for pest management. This study estimated the biomass losses resulting from pine wood nematode-caused *P. massoniana* diseases and included a correlation analysis with environmental factors on the sample scale, thereby leading to the following conclusions: 1) topographic factors and forest stand factors affect the *P. massoniana* biomass losses resulting from pine wood nematode-caused disease differently; 2) topographical factors have a greater impact on pest-related biomass losses caused by pine wood nematodes than forest stand factors; 3) the marginal effect diagram shows that elevation and slope were obviously related to biomass losses; biomass losses were positively related to elevation and negatively related to slope.

**List Of Abbreviations**
Declarations

Ethics approval and consent to participate

Not applicable

Consent for publication

Not applicable

Availability of data and material

The datasets used and/or analyzed during the current study are available from the corresponding author on reasonable request.

Competing interests

The authors declare that they have no competing interests.

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Authors' contributions

Zhiwei Wu and Shitao Lin designed the experiments; Zhibin Fang and Qiang Liu performed the experiments; Zhibin Fang analyzed the data and wrote the paper. All authors have read and approved the final manuscript.

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**Figures**
Figure 1

Location and sample point distribution map of Nankang District, Jiangxi Province Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

Figure 2

The overall steps of this study

Figure 3
Sample plot design

Figure 4

Multiple comparisons of biomass losses under different topographic and stand conditions (different letters in the box indicate significant differences, and the same letters indicate differences that are not significant)
Figure 5

The relative importance of each influencing factor in the random forest model
Figure 6

Marginal utility between Pinus thunbergii biomass loss rates and various influencing factors
Figure 7

Elevation and biomass loss in Nankang District Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.