

Estimation of the net primary productivity of forests on age in China

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Abstract. Under the background of global change, the internal relationship between forests' age and forests' carbon cycling is very important to study the position and function of forests in the global carbon cycle. Thus, the study employed field observations to inversely estimate the parameter of maximum light-use efficiency (ϵ_{\max}) of forests in different stand ages, generated a national map of forests' NPP on forests age. We found the ϵ_{\max} has decreased with the increased of forest ages that is consistent with other studies, and compared with the fixed ϵ_{\max} , the NPP of forests was calculated on forests age are more reliable and stable. The results shows a significant heterogeneity of forests' NPP_age, and the average NPP of the evergreen broadleaf forest, the deciduous broadleaf forest, evergreen needleleaf forest and deciduous needleleaf forests is about 377.34, 293.99, 273.63 and 306.36 gC/m²•yr. On the national level, the forests' NPP_age was 303.45 C/m²•yr in China. In the future, the forests' age needs to be taken into account in the Estimation of the net primary productivity.

INTRODUCTION

The growth status of forests is an important factor to determine the position of forests in the global carbon cycle [1]. Net primary productivity (NPP) not only directly reflects the forests in the natural environment conditions, the production capacity, characterization of forest growth [2], which also is the main index to establish the biological process model to predict and forecast the carbon cycle [3]. However, there is little understanding of the response of forests' growth to environmental change [4]. Under the background of global change, revealing the internal relationship between forests' age and forests' carbon cycling, it is very important to study the position and function of forests in the global carbon cycle [5, 6].

The remotely sensed forest NPP and NPP_obs retrieved from field observations were used, and a relationship between the maximum light-use efficiency and the forest stand ages was generated in the study. The forests age in this relationships as an intermediate parameter for major forest types in China.

DATA AND METHODS

DATA

(1) Remote sensing data. The Global Inventory Modeling and Mapping Studies (GIMMS) normalized difference vegetation index (NDVI) data sets (monthly, 8 km × 8 km, January 1982 to

December 2013) were generated to provide a 32-year satellite record of monthly changes in terrestrial vegetation^[7]. [ENREF 7](#)

(2) Field observation data of NPP. The field observation data of NPP contained extensive records of forest stands (10 m × 10 m plots) across China [8]. The database records included a total of 6153 forest stands.

(3) Meteorological data. The temperature and precipitation data were got from Climatic Research Unit (CRU) (0.5 ° × 0.5 °, 1982~2013). Total monthly solar radiation derived from 726 meteorological stations. This data were interpolated 0.0833 ° × 0.0833 ° used the Kriging method and a new interpolation base on digital elevation model.

(4) The vegetation data. The digitized vegetation map (1: 4,000,000) was acquired from Environmental and Ecological Science Data Center for West China, National Natural Science Foundation of China [9]. In the vector map, vegetation types in China were classified as follows: Evergreen broadleaf forest (EBF), Deciduous broadleaf forest (DBF), Evergreen needleleaf forest (ENF), Deciduous needleleaf forest (DNF), Mixed forest (MF).

(5) The forest stand ages data. The Spatial pattern of forests' ages in China with a spatial resolution of eight kilometer came from the Academy of Disaster Reduction and Emergency Management of Civil Affairs & Ministry of Education (ADREM) [10]. The average forests' age was 40.6 years in China, the main forests' age was ranged from 10 to 80 years.

Methods

CASA model

Carnegie-Ames-Stanford Approach (CASA) model [11] can estimate monthly NPP with satellite data, monthly temperature, precipitation and soil properties [12].

In this study, the CASA method of Zhu [13] was used to calculate vegetation NPP. In small scale of area, the NPP equals the amount of photosynthetically active radiation that absorbed by green vegetation (APAR) (g C · MJ⁻¹ · month⁻¹), multiplied by the actual light use efficiency (ε) (g C · MJ⁻¹), which the radiation is converted to plant biomass increment. Where x is a pixel of remote sensing image, and t is the period of NPP, such as a month.

$$NPP(x,t) = APAR(x,t) \times \varepsilon(x,t) \quad (1)$$

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Maximum light-use efficiency (ε_{max}) of forests' age in China

(1) The estimation of NPP on the CASA model is largely affected by the maximum light-use efficiency (ε_{max}). In this study, 4821 sampling plots that both have stand biomass and forests' age data were selected from 6153 sampling plots. These data were randomly divided into two groups: experimental group and validation group.

$$\varepsilon_{max} = \frac{NPP_{obs} \cdot \varepsilon_{Ts}}{NPP_{Ts}} \quad (2)$$

(2) In this study, the experimental group to correct the ε_{max} for all forest types, where

NPP_{obs} is the field observation data of NPP. The NPP_{obs} is separated by the forest types and the forest stand ages, the forest stand ages were separated into 23 groups (10 years interval classification, more than 220 years were divided into as group 23). Each group of data below 2 samples is not involved in the process.

NPP_{rs} is the estimated NPP based on the meteorological data and the remote sensing data (NDVI). ϵ_{rs} is the fixed value for all the forest types and all the forest age, and which value is 0.389. The value of ϵ_{max} is various from different forest types and the forest stand ages according to the NPP_{obs} group.

(3) According to the ϵ_{max} of the experimental group, the forest age-forest vegetation maximum NPP model was calculated and compared with the NPP data of ground observation in the validation group.

RESULTS and DISCUSSION

The relationship between the maximum light-use efficiency(ϵ_{max}) and the forest ages

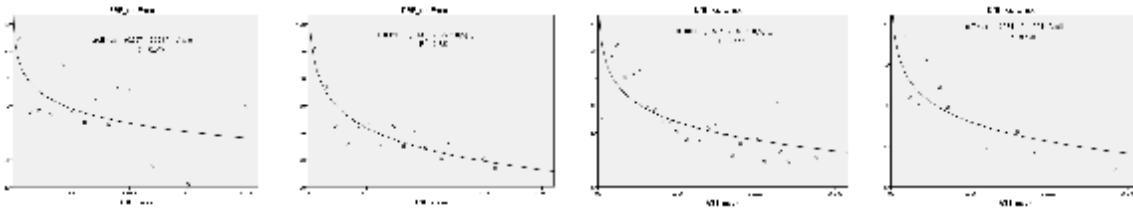


Fig.1. The relationship between the forest stand age and the ϵ_{max} of four forest types

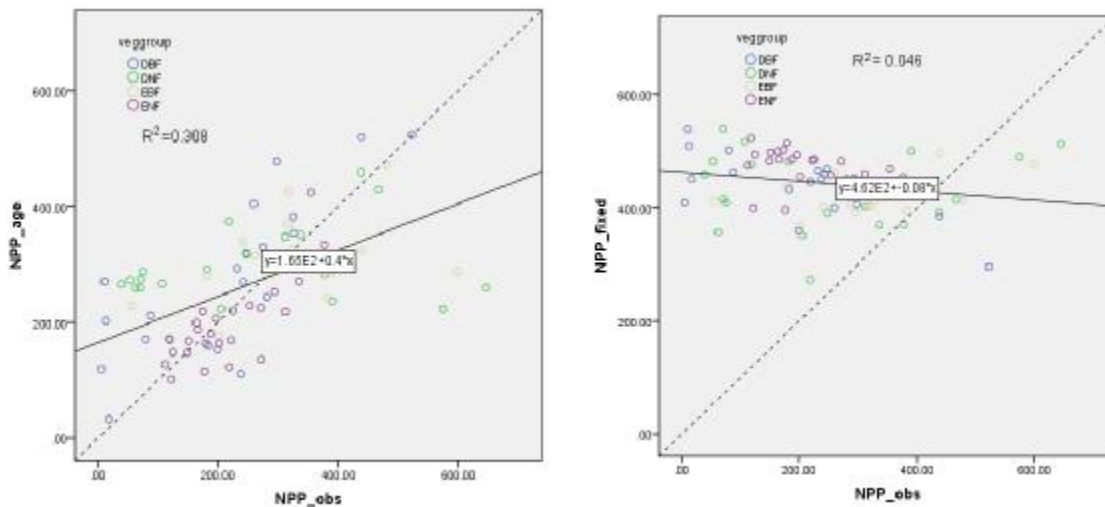


Fig.2. The relationship between NPP_{age} and NPP_{fixed} of four forest types.

The results show that the maximum light use efficiency was decreased with the increased of forest ages in EBF, DBF, ENF and DNF(Fig.1), which is consistent with the findings of Li, et al. [6, 14] . MF had little measured data in this study. Their (MF) was calculated with some data from references. The mean observed NPP of MF was 0.389. As shown in the Fig.2, the decisive coefficient R^2 between NPP_{age} and measured data NPP_{obs} increased from 0.046 to 0.308, which passed the significant test. The result show that the NPP of forests was calculated on forests age are reliable and stable.

The forests' NPP_{age} in China

The results indicate that the spatial distribution of forest NPP_{age} have significant heterogeneity. From the national level, the forest NPP_{age} was $303.45 C/m^2 \cdot yr$ in China. The forests' NPP_{age} over $500 C/m^2 \cdot yr$ are mainly distributed in central Hainan, southwestern Yunnan and south Tibet. The forests' NPP_{age} of the western Guangdong, central Guangxi, eastern Yunnan and south

Sichuan are mostly range from 350 to 500 $\text{gC/m}^2\cdot\text{yr}$. In the northeast China, deciduous broad-leaved forest are widely distributed in the large and small Xinganling Mountains and Changbai Mountain range, and the forest NPP value is mainly range from $200\text{gC/m}^2\cdot\text{yr}$ to $350\text{gC/m}^2\cdot\text{yr}$, partly in temperature and precipitation. There are still some forests' of NPP_age value are higher than $350 \text{gC/m}^2\cdot\text{yr}$, which because the photosynthetic efficiency of the forests is higher than other forests that with low age. In the central and eastern provinces of Hunan, Jiangxi, Fujian and Zhejiang, forest types are deciduous broad-leaved forests and deciduous coniferous forests, the forest NPP_age was 200 to $400 \text{gC/m}^2\cdot\text{yr}$ and the forests are young. The forest NPP_age of northwest China is lower than that of other regions in China, and the average values ranging from 5 to $15 \text{C/m}^2\cdot\text{yr}$ because low temperature and drought.

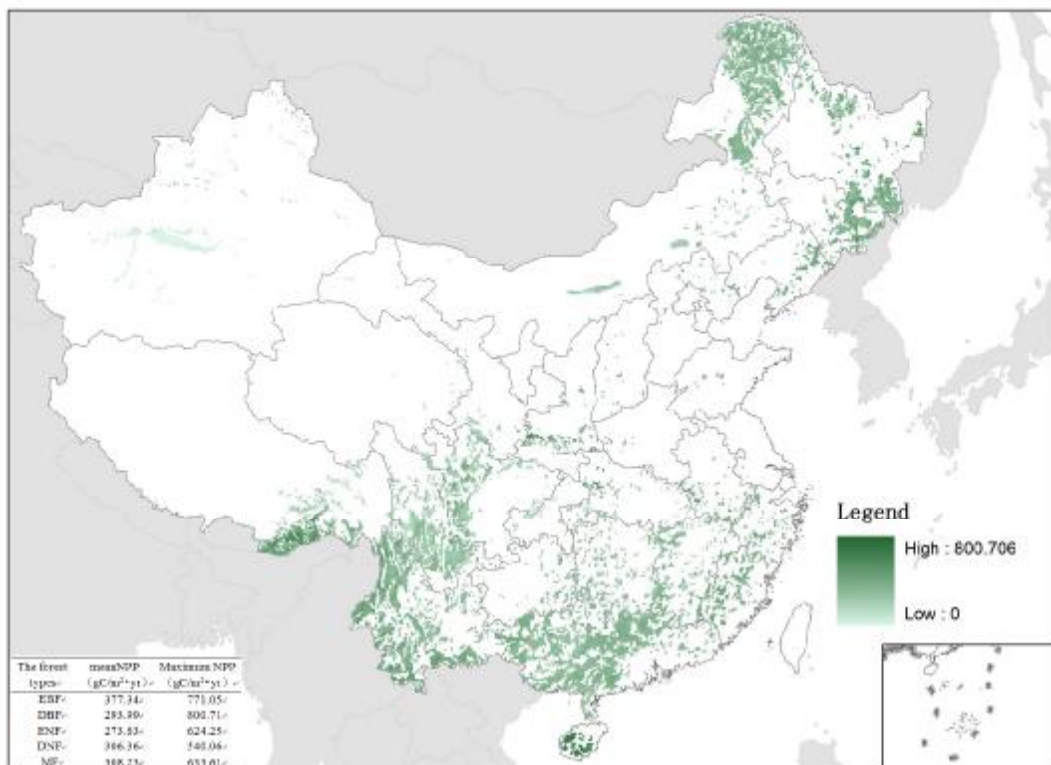


Fig.3. The distribution of NPP based on the forests age from 1982 to 2013

Estimated the NPP based on the forests agevary from different forest types. During the period 1982 to 2013, and the average NPP of the evergreen broadleaf forest, the deciduous broadleaf forest, evergreen needleleaf forest is about 377.34, 293.99 and $273.63 \text{gC/m}^2\cdot\text{yr}$, respectively, the average NPP of the evergreen needleleaf forests is less than each other. Compared with other forest types, the maximum forest NPP_age of DBF is the highest ($800.71 \text{gC/m}^2\cdot\text{yr}$) while the maximum forest NPP_age of DNF is the lowest ($540.06 \text{gC/m}^2\cdot\text{yr}$). Thus, the maximum forest NPP_age shows: $\text{DBF} > \text{EBF} > \text{ENF} > \text{MF} > \text{DNF}$.

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