Contrasting Effects of Temperature and Precipitation on Vegetation Greenness along Elevation Gradients of the Tibetan Plateau

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Abstract: The Tibetan Plateau (TP) is one of the most sensitive regions to global climate warming, not only at the inter-annual time scale but also at the altitudinal scale. We aim to investigate the contrasting effects of temperature and precipitation on vegetation greenness at different altitudes across the TP. In this study, interannual and elevational characteristics of the Normalized Difference Vegetation Index (NDVI), temperature, and precipitation were examined during the growing season from 1982 to 2015. We compared the elevational movement rates of the isolines of NDVI, temperature, and precipitation, and the sensitivities of elevational NDVI changes to temperature and precipitation. The results show that from 1982 to 2015, the elevational variation rate of isolines for NDVI mismatched with that for temperature and precipitation. The elevational movements of NDVI isolines were mostly controlled by precipitation at elevations below 2400 m and by the temperature at elevations above 2400 m. Precipitation appears to plays a role similar to temperature, and even a more effective role than the temperature at low elevations, in controlling elevational vegetation greenness changes at both spatial and interannual scales in the TP. This study highlights the regulation of temperature and precipitation on vegetation ecosystems along elevation gradients over the whole TP under global warming conditions.

Keywords: vegetation greenness; precipitation; temperature; elevational variation; sensitivity; isoline
1. Introduction

The Tibetan Plateau (TP) is known as the ‘roof of the world’ and ‘the Earth’s third pole’. It has an area of about 2.5 × 10^6 km^2, over half of which lies above 4000 m [1,2]. The TP is covered by a variety of vegetation up to a height of about 6000 m and can be divided into several vegetation ecosystems [3,4]. Numerous evidence indicate that the TP has been experiencing continuous climate warming during the past few decades, particularly since the 1980s [5–7]. Compared to the Northern Hemisphere, such warming started earlier [8,9] and is taking place faster than the global mean warming rate [10]. Warming-induced snowmelt causes the snowline to rise, and this process creates continuous positive feedback by stimulating the surface to receive more solar radiation [11–13]. The TP is characterized by arid and semi-arid climate [14,15]. Annual precipitation ranges from tens to hundreds of millimeters, with an overall increasing trend during the period 1979–2013 despite a distinct spatial variability [16,17]. Summer monsoons, westerlies, and local hydrological recycling are the water sources regulating moisture on the TP [10,18]. There is strong evidence that clear interactions exist between terrestrial ecosystems and climate variables [19–21], and vegetation greenness on the TP is highly sensitive to climate change [16,22]. Hydrothermal conditions play an essential role in the vegetation growth cycle [22,23]. Temperature is considered one of the most important factors that limit vegetation growth on the TP [24,25]. Previous studies indicated that a continuous increase in temperature leads to vegetation greenness amplitude changes on the TP [20,26,27], and also impacts the leaf-unfolding date [28,29]. For example, the nighttime temperature may play a key role in advancing greenup [30]. Higher air temperature could also indirectly result in vegetation growth reduction by increasing land surface evaporation and thus limiting water availability [31,32]. The role of temperature, however, may be overestimated according to recent research on ecosystem optimal air temperature, which suggested that colder regions have a lower optimal temperature [33]. Precipitation also affects spring phenology, and preseason precipitation can lead to earlier SOS (starting date of vegetation growing season) and reduce SOS sensitivity to preseason rainfall [34,35]. The vegetation of drier areas was found to be more dependent on precipitation [36,37]. Normalized Difference Vegetation Index (NDVI) dynamics influenced by precipitation on the northeastern and southwestern TP was also investigated by recent studies [23]. Fu et al. [38] found that the heat requirements of vegetation vary with precipitation availability, indicating the indirect effects of precipitation on vegetation growth. A comprehensive analysis was also conducted by Li et al. to quantify the contribution of temperature, precipitation, and solar radiation on vegetation growth [39]. Warming and wetting facilitate vegetation variation along elevation gradients in the alpine region [24,40]. Some studies have also described the effects of warmer temperature on biodiversity and ecological structure in alpine regions [41–43]. For example, elevation and slope influenced the distribution of vegetation cover while such an effect could be attributed to the variation of temperature and water availability [44,45]. Studies based on field and remote sensing also suggested the upward movement of species [46,47]. Pepin et al. has illustrated elevation-dependent warming in alpine areas [48].

Previous studies either did not take the synchronicity between the vegetation greenness variation and climate change into account or only analyzed the elevation-dependent relationship between temperature and greenness over 3000 m above sea level, thus excluded the role of precipitation and areas below 3000 m [24]. In addition, previous studies in other areas have found nonlinear and spatially heterogeneous interactions between temperature and greenness, implying that there could be a non-negligible effect of precipitation on greenness [35,49–52]. Therefore, it is necessary to find out how vegetation greenness responds to temperature and precipitation at different elevations over the whole TP under global warming conditions.

Because of the profound consequences for the terrestrial ecosystem [53], it is important to study the vegetation response to climate change over a long time scale [29,37,54]. In this study, Global Inventory Modeling and Mapping Studies (GIMMS) Normalized Difference Vegetation Index (NDVI) 3g data, which is the existing NDVI product with the longest time series and the finest spatial resolution, were used in combination with temperature and precipitation data from 1982 to 2015. We first explored
the synchronicity of the elevational variation in vegetation greenness and climate variables (i.e., temperature and precipitation). Then we investigated how vegetation greenness variation responds to climate change along elevation. There is no previous study on this subject that covered the entire TP over the period from 1982 to 2015. Our findings will provide insights for further understanding of the spatial changes of vegetation greenness along with altitude on the TP or other alpine regions affected by climate over the years.

2. Materials and Methods

2.1. Datasets

GIMMS NDVI3g (8 km) from 1982 to 2015 was downloaded from https://ecocast.arc.nasa.gov/data/pub/gimms/3g.v1/. The dataset was produced using 15-day maximum value composites with a spatial resolution of 8 km [55]. Moderate Resolution Imaging Spectroradiometer (MODIS) NDVI (250 m) from 2000 to 2015 extracted from the MOD13Q1 Version 6 product with a spatial resolution of 250m (https://reverb.echo.nasa.gov) was also obtained. MODIS NDVI data were produced by the constrained view angle-maximum value composite for each 16-day period. NDVI from GIMMS and MODIS were used to describe vegetation greenness. Precipitation and temperature from 1982 to 2015 were obtained from the China Meteorological Forcing Data (CMFD) of the National Tibetan Plateau Data Center (https://data.tpdc.ac.cn/en/data/). The precipitation and temperature data were produced by the fusion of remote sensing products, reanalysis datasets, and in situ observation data from weather stations. The temporal resolution of the CMFD is 3 h, and the horizontal spatial resolution is 0.1 degrees [56,57]. We also downloaded the WorldClim dataset (http://worldclim.org/version2), which includes monthly average temperature and precipitation data. Monthly average temperature (or precipitation) data has 12 bands (from January to December), each of which results from the average value of the corresponding month from 1970 to 2000. The WorldClim dataset was generated with data collected from between 9000 and 60,000 weather stations using the thin-plate splines method [58]. The ASTER (Advanced Spaceborne Thermal Emission and Reflected Radiometer) GDEM (Global Digital Elevation Model) V2 (30 m) was jointly released by NASA and the Ministry of Economy, Trade, and Industry of Japan (METI) can be downloaded from http://earthexplorer.usgs.gov/. Before the analysis, temperature and precipitation of CMFD were resampled by bilinear interpolation to match the spatial resolution of GIMMS NDVI3g [24]. The bilinear interpolation method averages the sum of the values of the four pixels (top, bottom, left, and right) adjacent to the target pixel as the output value of target pixel [59], which can provide satisfactory results for resampling of images with smooth texture while ensuring image quality [60]. Since we intended to reflect the integrated altitude in a GIMMS NDVI3g pixel, the pixel aggregate method was applied to resample the DEM data to the resolution of GIMMS NDVI3g by the pixel aggregate method [61]. The pixel aggregate method averages all of the pixel values that contribute to the output pixel.

2.2. Method

2.2.1. Elevation Dependence of the Rates of NDVI$_{GS}$, T$_{GS}$ and P$_{GS}$ Isoline

We calculated the annual mean NDVI over the TP. We excluded the pixels with a minimum annual mean NDVI value of less than 0.1, which were considered to be unvegetated (barren rock, sand, or snow) [62]. We then identified the pixels that had a 34-year mean NDVI value greater than 0.4 as the regions with medium or dense evergreen vegetation cover. These regions were included as part of the vegetation areas [24]. The following two criteria were then used to extract pixels corresponding to deciduous vegetation from the remaining pixels: (1) the peak value of seasonal NDVI occurred between June and August; and (2) the peak value of seasonal NDVI was at least 1.2 times the mean NDVI from January to March [24].
The vegetation growing season for the TP was considered to extend from May to September [24]. We calculated the mean NDVI (NDVI<sub>GS</sub>), mean temperature (T<sub>GS</sub>, unit: °C), and accumulated precipitation (P<sub>GS</sub>, unit: mm) for the growing season. In addition, we calculated the temporal trends of NDVI<sub>GS</sub> (T<sub>_NDVI</sub><sub>GS</sub>, unit: 1/year), T<sub>GS</sub> (T<sub>_T</sub><sub>GS</sub>, unit: °C/year), and P<sub>GS</sub> (T<sub>_P</sub><sub>GS</sub>, mm/year) for the period 1982–2015 for each pixel.

At the regional level, the TP was divided into 51 elevation intervals of 100 m in the range from 600 m to 5600 m. The altitude “h” defines an elevation interval from (h − 100) m to (h + 100) m. The first five and last two elevation sections were excluded because of very few valid NDVI pixels. The altitudinal gradient of NDVI<sub>GS</sub> (A<sub>_NDVI</sub><sub>GS</sub>, unit: °C/m), and P<sub>GS</sub> (A<sub>_P</sub><sub>GS</sub>) were then calculated for each elevation section. A<sub>_NDVI</sub><sub>GS</sub> was defined as:

\[
A_{NDVI_{GS}}(h) = \frac{NDVI_{GS}(h - 100 : h) - NDVI_{GS}(h : h + 100)}{E(h - 100 : h) - E(h : h + 100)}
\]

Where NDVI<sub>GS</sub> represents the mean value of the NDVI during the growing season for the elevation section h (unit: meters), E represents the elevation (unit: meters), and E(h−100: h) represents the mean elevation of all valid pixels in the elevation sections from h-100 to h. E(h: h + 100), NDVI<sub>GS</sub>(h−100: h), and NDVI<sub>GS</sub>(h: h + 100) are similarly defined.

The elevation variation rates of the NDVI<sub>GS</sub> (or P<sub>GS</sub>, T<sub>GS</sub>) isolines were defined as the annual change in the growing season NDVI (or P<sub>GS</sub>, T<sub>GS</sub>) within the given altitude section h. The elevation variation rate of the NDVI<sub>GS</sub> (or T<sub>GS</sub>, P<sub>GS</sub>) isolines, i.e., V<sub>_NDVI</sub><sub>GS</sub> (or V<sub>_T</sub><sub>GS</sub>, V<sub>_P</sub><sub>GS</sub>) (unit: m/year), was calculated by dividing T<sub>_NDVI</sub><sub>GS</sub> (or T<sub>_T</sub><sub>GS</sub>, T<sub>_P</sub><sub>GS</sub>) by A<sub>_NDVI</sub><sub>GS</sub> (or A<sub>_T</sub><sub>GS</sub>, A<sub>_P</sub><sub>GS</sub>). Therefore, the direction of the elevational movement of the NDVI<sub>GS</sub> isoline was jointly determined by T<sub>_NDVI</sub><sub>GS</sub>(h) and A<sub>_NDVI</sub><sub>GS</sub>(h).

2.2.2. Elevation Dependence of the Sensitivity of the NDVI<sub>GS</sub> Isoline Movement to Temperature and Precipitation

The sensitivity of the elevational motion of the NDVI<sub>GS</sub> isolines to temperature (S<sub>_NDVI</sub><sub>GS</sub>·T, unit: m/°C) and precipitation (S<sub>_NDVI</sub><sub>GS</sub>·P) were calculated for each elevation section:

\[
S_{NDVI_{GS}T}(h) = \frac{L_{NDVI_{GS}T}(h)}{A_{NDVI_{GS}}(h)}
\]

\[
S_{NDVI_{GS}P}(h) = \frac{L_{NDVI_{GS}P}(h)}{A_{NDVI_{GS}}(h)}
\]

where L<sub>_NDVI</sub><sub>GS</sub>·T(h) (unit: 1/°C) (or L<sub>_NDVI</sub><sub>GS</sub>·P(h), unit: 1/mm) is defined as the coefficient of T<sub>GS</sub>(h) (or P<sub>GS</sub>(h)) and obtained from a linear regression of NDVI<sub>GS</sub>(h) with T<sub>GS</sub>(h) and P<sub>GS</sub>(h). The direction of the sensitivity is determined by the coefficient of T<sub>GS</sub> (or P<sub>GS</sub>) in the regression equation and by A<sub>_NDVI</sub><sub>GS</sub>(h).

The elevational movement of the NDVI<sub>GS</sub>, T<sub>GS</sub>, and P<sub>GS</sub> isolines can be either upward or downward, with the direction being determined by the altitudinal gradient and temporal trend of the variables. Specifically, taking NDVI<sub>GS</sub> as an example, the upward movements of the NDVI<sub>GS</sub> isolines include: A mode (Figure 1a), i.e., a positive temporal trend (greenness increases with time) with a negative altitudinal gradient (greenness decreases with increasing altitude); and B mode (Figure 1b), i.e., a negative temporal trend (greenness decreases with time) with a positive altitudinal gradient (greenness increases with altitude). The downward movements of the NDVI<sub>GS</sub> isolines include: C mode, i.e., a negative temporal trend with a negative altitudinal gradient (Figure 1c); and D mode, i.e., a positive temporal trend with a positive altitudinal gradient (Figure 1d). This method can be used to determine if the direction of the sensitivity of the NDVI<sub>GS</sub> isolines to temperature and precipitation is upward or downward.
At the pixel level, we used a 3 × 3 moving window to find the altitudinal gradient of the central pixels. Specifically, a 3 × 3-pixels region can be divided into four directions through the central pixel: horizontal, vertical, and two oblique directions. The direction was retained if the elevation increased or decreased monotonically from one side to another in this direction. We then selected the direction with the largest elevation difference from the remaining directions as the final direction and calculated the altitudinal gradient of the central pixel using the altitude value of the three pixels lying in this final selected direction. Then, the elevational variation rate of the NDVIGS, TGS, and PGS isolines was calculated for the central pixel as well as the sensitivity of the isoline elevational movement to temperature and precipitation.

Finally, the altitudinal gradient and elevational movement of the isolines for temperature and precipitation derived from the CMFD and the WorldClim dataset were calculated. We also compared the sensitivities of the elevational movement of the NDVIGS isolines to temperature and precipitation using both 250-m MODIS NDVI and 8-km GIMMS NDVI3g to examine the responses of vegetation greenness to climate changes using imagery with two different spatial resolutions.
3. Results

3.1. Dependence of the Rate of NDVI$_{GS}$, T$_{GS}$, and P$_{GS}$ Isolines on Elevation

The elevational movement of the vegetation greenness isolines mismatched with that of the temperature and precipitation isolines (Figure 2). The elevational variation rate of the T$_{GS}$ isoline was positive for almost the whole TP whereas the rate of NDVI$_{GS}$ and P$_{GS}$ isoline were both positive and negative (Figure 2). In addition, the value of the NDVI$_{GS}$ isoline variation rate along the elevation was much smaller than that of T$_{GS}$ isoline and P$_{GS}$ isoline. For most areas, the elevational variation rate of the NDVI$_{GS}$ isoline was concentrated in the intervals of 0-10 m/year and −10-0 m/year, accounting for 37% and 33% of the TP, respectively (Figure 2a). Pixels with the NDVI$_{GS}$ isoline motion rate of >20 m/year or <−20 m/year accounted for only 9% and 8% of the TP, respectively. The elevational variation rate of the NDVI$_{GS}$ isoline was higher at lower altitudes than that at higher altitudes. There were upward and downward movements of the NDVI$_{GS}$ isolines across the TP (Figure 2b), with the movement being upward for 53% of the TP. The temporal trend of NDVI$_{GS}$ and the gradient of NDVI$_{GS}$ with altitude caused either upward or downward motion of the NDVI$_{GS}$ isolines. Our results show that 67% of all NDVI pixels decreased with altitude, and 37% of these NDVI$_{GS}$ isolines moved upward (A in Figure 2b) while 30% moved downward (C in Figure 2b). The remaining 33% of the area of the TP had positive gradients of NDVI$_{GS}$ with altitude, 16% of which were upward because NDVI$_{GS}$ decreased. V$_{T_{GS}}$ on the TP was mainly in the range of 0-20 m/year (Figure 2c). Overall, The T$_{GS}$ isolines mainly moved upward, accounting for about 90% of all pixels (A in Figure 2d). The upward movement of the T$_{GS}$ isolines was due to the negative gradient of T$_{GS}$ with altitude and the positive temporal trend of T$_{GS}$. The elevational variation rate of the P$_{GS}$ isoline was within the ranges of −50–0 (21%) and 0 50 m/year (26%) (Figure 2e). The P$_{GS}$ isolines in 47% of all the pixels moved upward due to the negative altitudinal gradient of P$_{GS}$, and P$_{GS}$ isolines moved downwards along with the positive altitudinal gradient in about 40% pixels (Figure 2f).
At the regional level, a mismatch was also found between the elevational movement of the vegetation greenness isolines and those of the temperature and precipitation isolines. The elevational variation rate of the $V_{\text{NDVI GS}}$ isolines was smaller than the rate of $V_{\text{T GS}}$ and $V_{\text{P GS}}$ isolines (Figure 3). The elevational variation rate of the $V_{\text{NDVI GS}}$ isolines was greater at low altitudes (below 2400 m) than that at high altitudes with strong fluctuations (Figure 3). The elevational variation rate of the $V_{\text{NDVI GS}}$ isoline varied gently from 2500 m to 5600 m. Compared with the variation at low altitudes, the value of the elevational variation rate (between 1 m/year and 10 m/year) from 2500 m to 4100 m was an order of magnitude lower than that below 2400 m. Above 4100 m, the value was an order of magnitude lower than in the 2500–4100 m interval and remained between $-1$ m/year and 1 m/year. The movement of the $V_{\text{T GS}}$ isolines was uniformly upward, with the minimum variation rate of the $V_{\text{T GS}}$ isoline at 3100 m (3.85 m/year). Above that, the elevational variation rate of $V_{\text{T GS}}$ increased with two peaks of 54 m/year at 3900 m and 37 m/year at 4700 m and decreased to 22 m/year at 5600 m. The elevational variation
rate of the $P_{GS}$ isolines fluctuated greatly between 600 m and 1700 m. The maximum negative rate was about $-300 \, \text{m/year}$ at 1700 m and the maximum positive rate was about $300 \, \text{m/year}$ at 1500 m. Above 1700 m, except for the 3900 m, 5000 m, and 5100 m sections, the rate was about $10 \, \text{m/year}$.

Figure 3. Dependence upon elevation of the elevational variation rate of NDVICGS isoline ($V_{NDVICGS}$, black solid line), $T_{GS}$ isoline ($V_{T_{GS}}$, red solid line), and $P_{GS}$ isoline ($V_{P_{GS}}$, blue solid line) plotted against elevation.

3.2. Sensitivity of the Movement of the NDVICGS Isolines to Temperature and Precipitation

The sensitivity of the elevational movement of the NDVICGS isolines to temperature was higher at low altitudes than at high altitudes (Figure 4a), whereas this trend was not obvious for the sensitivity to precipitation. The sensitivity to temperature was within the interval of $-50$–$0 \, \degree \text{C}$ for 22% of all the pixels and within the interval of $0$–$50 \, \degree \text{C}$ for 27% of all the pixels. For most pixels, the sensitivity to precipitation was within the ranges of $-1$–$0 \, \text{m/mm}$ (about 39%) and $0$–$1 \, \text{m/mm}$ (for about 33%) (Figure 4c). The NDVICGS isolines had both positive and negative responses to changes in precipitation and temperature. Specifically, the NDVICGS isolines responded positively to temperature increase (i.e., moved upward) for about 54% of the area and negatively for the remaining 46% of the area (Figure 4b). The expected positive responses to temperature occurred in the eastern TP whereas negative responses occurred in the southern TP (Figure 4a). Precipitation had an opposite effect on the movement of the NDVICGS isolines. About 54% of the NDVICGS isolines moved downward in response to increased precipitation (Figure 4d).
Figure 4. Distributions of (a) the sensitivity of the NDVI<sub>GS</sub> isolines elevational movements to temperature (S<sub>NDVI<sub>GS</sub>_T)), (b) the direction of the sensitivity of the NDVI isolines elevational movement to temperature, (c) the sensitivity of NDVI<sub>GS</sub> isolines elevational movement to precipitation (S<sub>NDVI<sub>GS</sub>_P)), and (d) the direction of the sensitivity of the NDVI isolines elevational movement to precipitation. A, B, C, and D modes in (b,d) are defined in Figure 1. The bar charts in the lower-left corner of each graph show the percentage of pixels in each interval.

Below 2400 m, the elevational variation rate of the NDVI<sub>GS</sub> isoline and the sensitivity of the NDVI<sub>GS</sub> isoline elevational variation to precipitation were both positive (or negative) (Figure 5b). Above 2500 m, the elevational variation rate of the NDVI<sub>GS</sub> isoline and the sensitivity of the NDVI<sub>GS</sub> isoline elevational variation to temperature were both positive or negative (Figure 5a). The values of the sensitivity to temperature were generally within the range between $-193$ m/°C and 144 m/°C below 3600 m, which was larger than the values above 3700 m (Figure 5a). The values of the sensitivity to precipitation were between $-1$ m/mm and 1 m/mm except for the values of $-4.6$ m/mm at 1400 m, 3.4 m/mm at 1800 m, and 2.0 m/mm at 2300 m (Figure 5b).
Figure 5. (a) The sensitivity of the elevational movement of the NDVI_S isolines to temperature ($S_{NDVI_S-T}$, red solid line) plotted along with the elevational variation rate of NDVI_S ($V_{NDVI_S}$, black solid line) against the elevation; (b) the sensitivity of the elevational movement of the NDVI_S isolines to precipitation ($S_{NDVI_S-P}$, blue solid line) plotted along with the elevational variation rate of NDVI_S (black solid line) against the elevation.

3.3. The Responses of Vegetation Greenness to Climate at Different Spatial Scales

We compared the elevational variation rates of temperature and precipitation isolines to investigate the uncertainties caused by differences in spatial resolutions, i.e., 0.1-degrees for CMFD and 1-km for WorldClim (Figure 6). The results show that the rates obtained from the two climate datasets were highly consistent in space, and about 80% elevational variation rates of $T_{GS}$ isolines were concentrated in the range of 0–5 m/year (Figure 6a,b). More heterogeneities can be observed for the spatial distribution of the elevational variation rate of the $P_{GS}$ isolines (Figure 6c,d). However, there was also consistency between the two sets of elevational variation rates of the $P_{GS}$ isolines.

Figure 6. Spatial patterns of elevational variation rate of (a) temperature isolines with a spatial resolution of 0.1 degrees, (b) temperature isolines with a spatial resolution of 1 km, (c) precipitation isolines with a spatial resolution of 0.1 degrees, and (d) precipitation isolines with a spatial resolution of 1 km.
We used 8-km GIMMS NDVI3g and 250-m MODIS NDVI from 2000 to 2015 to calculate the elevational movement of the NDVI\textsubscript{GS} isolines and then examined their sensitivities to temperature and precipitation. Significant correlations were found for their sensitivities to temperature \((r = 0.52, p < 0.001)\) (Figure 7a) and to precipitation \((r = 0.84, p < 0.001)\) (Figure 7b). Our results indicate that 8-km GIMMS NDVI showed similar elevational contrasting effects of temperature and precipitation across the TP compared with 250-m MODIS NDVI. Moreover, GIMMS NDVI, taking advantage of a longer time series (from 1982 to 2015) than MODIS NDVI, is more useful to explore long-term vegetation responses to climate change.

![Figure 7. Correlation between the sensitivities of the elevational movements of GIMMS NDVI\textsubscript{GS} and MODIS NDVI\textsubscript{GS} to temperature (S\textsubscript{NDVI\textsubscript{GS}_T} and S\textsubscript{MODIS NDVI\textsubscript{GS}_T}) (a) and to precipitation (S\textsubscript{NDVI\textsubscript{GS}_P} and S\textsubscript{MODIS NDVI\textsubscript{GS}_P}) (b). *** indicates the significant correlation at the level of \(p < 0.001\).](image)

4. Discussion

Climate warming has caused upward movements of growing season temperature isolines; growing season precipitation isolines have also moved upward over about half of the TP. However, the elevational movement of the vegetation greenness isolines does not always match the isolate movements of temperature and precipitation. The rate of change of the vegetation greenness isolines is lower than that of the temperature and precipitation isolines (Figure 3). There may be several reasons for this: first, the elevational movement of the growing season temperature and precipitation isolines is asynchronous. The availability of heat and water changes with increasing altitude and the combined effect of these also changes [50,63]. High temperatures may have little positive effect on vegetation when water availability is limited [20,35,49,64]. This may also cause a slow movement of the temperature isolines and a rapid movement of the precipitation isolines [65]. A previous study showed that the warming rate is faster at high altitudes [48], which might result in faster evaporation in these areas and cause a hydrothermal imbalance due to temperature and water availability. The amount of precipitation does not equate to the amount of water available to vegetation on the TP, which mostly consists of high mountains. Particularly, steep slopes can affect water storage and cause water loss, meaning that water cannot effectively be used by vegetation when the ground is saturated [44,65]. Differences in aspect also affect the local humidity due to differences in the duration of sunshine. Terrain further affects the distribution of vegetation types because the forest is often found in areas with the steepest slopes while meadows are concentrated in flatter areas. Another point to consider is that while accelerating warming can alleviate the environmental constraints on vegetation growth, vegetation will also concurrently require more nutrients [39,66–68]. Our results showed that there was a continuous increase in temperature over the TP during the study period and that precipitation also increased. As a result, vegetation’s need for nutrients such as nitrogen and CO\textsubscript{2} may have increased,
faster than they were supplied. It has also been shown that the start of the growing season on the TP has become earlier [28,29]. In some areas of the TP, moreover, the date of autumn leaf senescence was delayed more significantly than the spring green-up date was anticipated [69]. Although these phenomena may lead to a prolonged growing season, it does not necessarily mean that vegetation becomes greener. Rapid growth early in the growing season may be offset by a rapid decline at the end due to environmental limitations (e.g., water stress) [70]. Vegetation may also become more tolerant of warmer and wetter conditions [71,72].

Human activities may have an impact on vegetation growth. The decline in grazing on the alpine grasslands of the TP may have offset the negative impacts of climate change over the last decade [73]. Conservation efforts such as the Three-North Shelterbelt Project, The Natural Forest Protection Project, and Program to Return Farmland to Forests also contributed to vegetation recovery, especially in the northeast of the TP [45]. All these facts indicate that anthropogenic activities have a significant effect on alpine ecosystems [54].

Previous studies have shown that interactions among species can postpone the upward movement of treelines and shrub lines on the TP. For example, treeline establishment under warming condition is controlled by many abiotic factors (e.g., temperature, precipitation, human disturbance) and biotic factors (e.g., mean root-zone, proximity to trees, canopy cover), and the importance of these factors may fluctuate under different conditions [74–78]. Biotic interspecific competition between the tree population and its upper-elevation population (e.g., shrub or grass) cannot be ignored when the treeline varies along elevation [77,79,80]. The denser the shrub population, the greater the restrictions imposed on treeline establishment [40]. Upper vegetation, however, also provides an ameliorated biotic microenvironment for seedlings and saplings growth [81]. Proximity to a tree is also a significant biotic factor, which maintains the biological requirement or prevents seedlings or saplings from frost or more Ultraviolet-B radiation due to the existence of understory vegetation cover or canopy cover [75,77,82,83].

These earlier studies provided insights into how the composition of species communities affects the elevational movement of the vegetation greenness isolines [40], and into how climate change may have different effects on species composition and vegetation at different altitudes [84,85]. The TP plays an important role in controlling environmental changes in surrounding regions and across the northern hemisphere as a whole [86]. Many abiotic and biotic factors affect alpine vegetation. Therefore, future studies of the variation in vegetation greenness along the elevational gradients of the TP should integrate external environmental factors with the biophysical processes regulating vegetation growth.

Our results suggest that the temperature and precipitation together regulate the greening of vegetation on the TP, indicating that the elevational variation in vegetation greenness isolines over the years were influenced more by precipitation at elevations below 2400 m (Figure 5b), while above 2400 m, it depended more on temperature (Figure 5a). Such a phenomenon is consistent with Zhang et al. [45], who suggested that temperature is the dominant control on the vegetation fractional coverage at a higher elevation. The transitional elevation zone from precipitation dominated to temperature dominated behaved differently. Such discrepancy may be because of differences in the study period and in the elevational intervals. Changes in climatic conditions caused by latitudinal differences also cause changes in the main factors limiting vegetation growth [87]. Such effect may exist in some alpine regions [88] where vegetation greening might be more sensitive to precipitation at lower altitudes in areas where the temperature is relatively moderate. At higher altitudes, i.e., the colder regions, a small increase in temperature can produce a relatively large increase in the thermal balance [45,89]. Therefore, vegetation may be more sensitive to temperatures at high altitudes once the limitations caused by low temperatures are removed. During the growing season, the ratio of accumulated precipitation to ≥5 °C accumulated temperature varies with elevation, and this ratio determines which factor (precipitation or temperature) is more important to vegetation growth [88,90], indicating that the regulation of vegetation growth is jointly affected by precipitation and temperature. Vegetation types are another factor affecting the response of the vegetation greenness isolines to climate variables. For example, forests need a lot of rain whereas grasslands are more drought-tolerant. Generally, the TP
exhibits a vegetation gradient from forest and meadow to steppe and desert corresponding as altitude increases, and the sensitivity of different vegetation types to temperature and precipitation also varies with elevation [3,24,34].

5. Conclusions

In this study, we used a 34-year GIMMS NDVI3g dataset and temperature and precipitation data for the same period to analyze the elevational movement of vegetation isolines from 1982 to 2015 on the TP. Besides the influence of temperature on vegetation greening, we also highlighted the significant impact of precipitation on vegetation growth. Our main conclusions are: (1) vegetation greenness on the TP was controlled jointly by temperature and precipitation. Below 2400 m, the elevational movement of the vegetation greenness isolines was controlled more by precipitation than by temperature, and above 2400 m, it was controlled more by temperature than by precipitation. (2) Mismatched elevation variation rates exist between the vegetation greenness isolines and the temperature and precipitation isolines. There are many abiotic and biotic factors that might affect the elevational movement of the vegetation greenness isolines. This mismatch may lead to changes in the structure of the ecosystem of the TP. Given that the temperature on the Tibetan Plateau will continue to increase in the future, we believe that precipitation will still dominate vegetation greenness variation in low-altitude areas, and the alleviated temperature pressure in high-altitude areas will continue to promote vegetation greening in these regions. These findings will benefit ecological conservation efforts on the Tibetan Plateau.

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