Land Management Contributes significantly to observed Vegetation Browning in Syria during 2001–2018

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Abstract. Climate change and human activities have significant impacts on terrestrial vegetation. Syria is a typical arid region with a water-limited ecosystem and has experienced severe social unrest over the last decades. In this study, changes in vegetation and potential drivers in Syria are investigated. By using an enhanced vegetation index (EVI), a general browning trend is found in Syria during 2001–2018 with the EVI decreasing at a rate of −0.8×10⁻³ yr⁻¹ (p < 0.1). The decrease of the EVI is mainly found in the north region, whereas the west region still maintains an increasing trend. The residual analysis indicates that besides precipitation, human activities also contribute significantly to the EVI decrease, which is confirmed by the decrease in rainfall use efficiency. In order to reach a more reliable conclusion, a comparative analysis is carried out in the Khabur River Basin where croplands are widely distributed in adjacent regions of Syria and Turkey. The time series of the EVIs over these two regions are highly correlated (r = 0.8027, p < 0.001), indicating that both regions are affected by similar climate forcing. However, vegetation in Syria and Turkey illustrates contrary browning (−3×10⁻³ yr⁻¹, p < 0.01) and greening trends (4.5×10⁻³ yr⁻¹, p < 0.01), respectively. Relative reports have reported that social unrest had induced insufficient irrigation and lack of seeds, fertilizers, pesticides and field managements. Therefore, we concluded that the decline of vegetation in the north Syria is driven by the change of land managements.
1. Introduction

Vegetation is a key component of the Earth system and plays an important role in the water and energy cycle. It is also one of the basic natural conditions for human survival. Therefore, the variations, driving mechanisms and impacts of the vegetation have been widely studied (Li et al., 2018; Piao et al., 2020; Zhu et al., 2016). Globally, observations have shown a general greening trend, although browning, i.e. loss of vegetation activity occurred in some regions (Jong et al., 2012; Zhang et al., 2017). Here, the greening (browning) usually refers to changes in vegetation conditions indicated by the positive (negative) long-term trend of vegetation index (Piao et al., 2020).

Recent advances in remote sensing technology and the establishment of long-term vegetation indexes have greatly facilitated the research on the long-term monitoring of large-scale vegetation. Indexes such as the normalized difference vegetation index (NDVI) developed by the Global Inventory Monitoring and Modeling Studies (GIMMS), the enhanced vegetation index (EVI) product based on the Moderate Resolution Imaging Spectroradiometer (MODIS) and the leaf area index (LAI) derived from remote sensing data are widely used in the study of long-term vegetation changes (Forzieri et al., 2017; Ju and Masek, 2016). The main climatic factors restricting vegetation include temperature, precipitation, radiation, and the fertilization effect of CO$_2$ (Chen et al., 2020; Keenan and Riley, 2018; Schimel et al., 2015). At the global scale, the long-term trends of these driving factors manifest itself as warming and increasing atmospheric CO$_2$ concentration. Model results show that the CO$_2$ fertilization effect on global scale greening contributes over a half (Zhu et al., 2016). However, the vegetation index derived from remote sensing data shows that the proportion of global farmland greening is relatively high, which is different from model simulations on the CO$_2$ fertilization effect. At the same time, the agricultural modernization is also considered to have a great impact on vegetation growth (Chen et al., 2019).

At present, the influences of climate change on vegetation have been widely simulated by models, but the model performance on the variation of land-use types is relatively poor, and land management practices are not well represented (Pongratz et al., 2014; Prestele et al., 2017). There are two main reasons for these disadvantages. On one hand, the land management data are not complete. Land management involves many aspects, including ecological engineering, economic management of natural vegetation, as well as the seed-selection, irrigation, fertilization and pesticides for farmlands. This results in the lack of high-quality datasets with spatio-temporal continuity for land management. On the other hand, it is more difficult to quantify the characteristics of various types of land management. Therefore, a common practice in previous studies is to make a general distinction between natural and human drivers. As a result, it is rather difficult to study the specific management processes in depth (d’Amour et al., 2017; Erb et al., 2017; Meyfroidt et al., 2013).

From a land management perspective, Syria deserves great concern. Syria is a typical arid area, the social unrest in Syria is bound to lead to a series of changes in land management, especially the serious destruction to agricultural input caused by the war. So far, there have been relatively few studies on land management changes triggered by social unrest. A typical case in the past is the problem of cropland abandonment in Eastern Europe caused by the collapse of the Soviet Union (Alcantara et al., 2013; Schierhorn et al., 2013). Therefore, it is necessary for Syria to carefully study the impact and mechanism of
climate and human activities on vegetation changes. This is also the main research objective of this paper. Specifically, in this paper the basic characteristics of vegetation change in Syria are investigated. Furthermore, the impact of land management on vegetation changes is analyzed from multiple ways, so as to verify the hypothesis that the social unrest will significantly affect the vegetation. Although this mechanism is intuitively obvious, it is not easy to obtain effective and definite quantified conclusions. The remainder of this paper is organized as follows. Section 2 summarizes data and methods. Detailed results are presented in section 3. Discussions and conclusions are presented in section 4 and section 5, respectively.

2. Data and Methods

2.1 Study Area

Figure 1. Land cover and vegetation condition in Syria. (a) Land cover types based on the MOD12C1 data of 2010. Color codes 1–4 indicate the natural vegetation, croplands, water bodies and built-up lands, respectively. 0 represents the unused land which is colored in gray. (b) The spatial distribution of annual EVI in Syria.

Syria (Figure 1) is located in Western Asia, on the east bank of the Mediterranean Sea, bordering Turkey in the north, Iraq in the east, Jordan in the south, and Lebanon and Palestine in the southwest. Syria has a dry climate, and the precipitation is mainly concentrated in the winter half year (November to the next April). The Mediterranean climate in its coastal and northern parts leads to hot, dry summers but mild, rainy winters in these regions. In contrast, the southern region of Syria belongs to tropical desert climate with high temperature and drought throughout the whole year with less precipitation. The main vegetation types in Syria are farmland and grassland, which are primarily distributed in the northern and western coastal areas. The main crop grown in Syria is wheat (Mohammed et al., 2020).
2.2 Land Cover Type Data

In this paper, the land use and land cover dataset of MOD12C1 product from MODIS is used, with the spatial and temporal resolutions being 0.05°×0.05° and one year, respectively. This dataset has three different classification schemes. In this study, the IGBP scheme is selected and the data is re-grouped into five categories as shown in Figure 1(a). The natural vegetation includes evergreen needleleaf forests, evergreen broadleaf forests, deciduous needleleaf forests, deciduous broadleaf forests, mixed forests, open shrublands, woody savannas, savannas and grasslands. The second category is cropland that includes croplands and the mosaics of croplands and natural vegetation. The third category is water which contains permanent wetlands and water bodies. The built-up lands remain consistent with the original classification. The rest is classified as the fifth category, unused land, including barren and non-vegetated lands.

2.3 Vegetation Index

Optical vegetation indexes are widely used to study the vegetation state. In this paper, the EVI based on MODIS, which can be calculated according to Formula (1), is selected to characterize the vegetation. Compared with the NDVI, the EVI includes the information of the blue band, which can solve the saturation problem of NDVI in densely vegetated areas to a certain extent. The MOD13C2 product is selected, with its spatial resolution (0.05°×0.05°) being consistent with that of the land-use data. The time period of 2001–2018 is selected for the EVI calculation. In October 2018, unusually heavy rain hit Syria and other Middle East regions and caused flooding, which also led to a very high EVI in the following year of 2019. Meanwhile, the heavy rainfall also contributed to desert locust crisis in 2020 (Salih et al., 2020). Because the abnormal value in 2019 has a great interference on trend analysis, it is excluded in this paper. EVI can be calculated as follows:

\[
EVI = \frac{G \times (\rho_{\text{NIR}} - \rho_{\text{red}})}{\rho_{\text{NIR}} + \rho_{\text{red}} - \rho_{\text{red}} - \rho_{\text{blue}} + L}
\]

(1)

where \(\rho_{\text{NIR}}, \rho_{\text{red}}, \rho_{\text{blue}}\) are the reflectance of the near-infrared, red and blue bands, respectively; \(G (=2.5)\) is the gain factor; \(C_1 (=6), C_2 (=7.5)\) and \(L (=1)\) are adopted coefficients, respectively.

2.4 Climatic Factors

The main climatic factors affecting vegetation changes include radiation, precipitation and temperature. Due to the high temperature and drought in Syria, the moisture content is the main restrictive factor. At the same time, temperature has a huge impact on soil water loss through evaporation. Therefore, precipitation and temperature are selected as environmental factors. The CRU TS4.03 (Brown et al., 2020) dataset with the temporal resolution of one month and spatial resolution of 0.5°×0.5° is adopted in this paper. Comparatively, the spatial resolutions of temperature and precipitation are much lower than that of vegetation. However, due to improved spatial consistencies of temperature and precipitation, it is assumed that the vegetation within the 0.5° interval has the same environmental condition.
2.5 Statistical Methods

The residual trend analysis (Burrell et al., 2017; Evans and Geerken, 2004) is widely used to deduce the impact of human activities by removing climatic factor contributions. Its basic assumption is as follows. Through the regression of vegetation and climatic factors, the trend of residuals is analyzed. For a significant trend of residuals, it is considered that this trend is caused by other driving factors other than climatic factors, which is generally attributed to human activities.

The variation of rainfall use efficiency (RUE) is also analyzed, which is expressed by the ratio of EVI to precipitation (Fensholt and Rasmussen, 2011; Ibrahim et al., 2015). Previous studies have shown that there is a good linear relationship between above ground biomass and accumulated precipitation in semi-arid water-limited ecosystems. At the same time, the accumulative remote sensing vegetation index can well represent the aboveground net primary productivity (ANPP).

Changes in the RUE may be attributed to the factors other than water conditions, which can be generally considered as the influence of human activities (Leroux et al., 2017). The Pearson correlation analysis and Mann-Kendall trend analysis are used to analyze the correlation and trends, respectively.

3. Results

3.1 Seasonal Cycle and Long-term Trends
Figure 2. monthly variations of climatic factors and vegetation of Syria. (a) – (b) Precipitation (mm) and temperature (°C). (c) – (d) EVIs of natural vegetation and croplands. All the data are the averages during 2001–2018.

Figure 2 shows the monthly variations of the multi-year averaged hydrometeorological conditions (precipitation and temperature) and vegetation conditions in Syria. During the hot, dry summer in this region, there is little precipitation from June to September, with the total precipitation being only about 8.2 mm during this period. Rainfall begins to increase in October, reaching its peak in January, and then decreases again month by month. The monthly average temperature in winter is lower than 10°C. Since October, although the dry season has gradually turned to the rainy season, the vegetation index is still very low, including both farmland vegetation and natural vegetation (Figure 2(c) – (d)). The growing period of farmland and natural vegetation is concentrated in February to May, and the growth of farmland vegetation is obviously better than that of natural vegetation, which is reflected in higher EVI. Therefore, the period from February to May is generally selected as the growing period.

The inter-annual variations of precipitation, temperature and vegetation are further analyzed in Figure 3. It is found that in the whole year or in the growing period, both the farmland vegetation and the natural vegetation show a decreasing trend. Meanwhile, there is an increasing trend of temperature and a decreasing trend of precipitation. Therefore, intuitively, increases in temperature and decreases in precipitation enhance water constraints, leading to further deterioration of hydrometeorological conditions in this region.

The monthly variation of climatic factors and vegetation are consistent is the inter-annual variation. Particularly, the temperature only slightly decreases in October while it increases in other months. The most obvious decrease of precipitation
occurs in February (Figure 4(b)), with a rate of −2.14 mm yr⁻¹. However, vegetation in April and May decreases most significantly, including both farmland vegetation and natural vegetation. This can be most likely attributed to the precipitation decline in spring, especially the February.

Figure 3. Inter-annual variations of climatic factors and vegetation during 2001–2018. (a) – (d) are annual averages of temperature, precipitation, cropland EVI and natural vegetation EVI. And (e) – (h) are growing season averages of temperature, precipitation, cropland EVI and natural vegetation EVI.
3.2 Spatial Patterns of Long-term Trends

There are large differences in the spatial variation of the vegetation in the study area. As shown in Figure 5, the EVI in the growing period shows an increasing trend in the western area, while the decreasing trend is mainly concentrated in the north (northwest and northeast), especially for the farmland vegetation. The greening area in the south is also associated with decreases in precipitation and increases in temperature, which presents a problem in explaining this spatial distribution pattern through changes of temperature and water. Thus, it is necessary to further analyze this spatial pattern by residual analysis.
The vegetation change in the study area is mainly constrained by the soil moisture and precipitation. Therefore, the residual trend after removing the influence of precipitation is analyzed in this paper. Compared with overall precipitation, most of the water used by vegetation is soil moisture in root layer, which experiences cumulative and hysteresis effects (Chen et al., 2014). It is found that the EVI in the growing period is most sensitive to the cumulative precipitation from November of the previous year to April of this year (N-A precipitation for short). As shown in Figure 6, for most regions there are significant positive correlations between the N-A precipitation and the EVI in the growing period. Figure 6(b) shows the corresponding trends of residual changes. The significantly decreasing trends of the EVI residuals are mainly concentrated in northern Syria. The RUE here is expressed by the ratio of the EVI in the growing period to the N-A precipitation, of which the interannual variations also could be used to induce human activity impacts. The result shows that the RUE decreases in the north and increases in the southwest (Figure 6(c)), which is similar to that of residual analysis in spatial distribution. Therefore, based on the We suggest that in addition to climatic factors, the most important reason for vegetation decline in this area are changes in human activities.
Figure 6. (a) correlations between the N-A precipitation and the EVI in the growing period; (b) trends of EVI residuals after removing the linear fitting by precipitation; (c) rainfall use efficiency.

3.3 Paired analysis

In order to illustrate the land management rather than climate impact on vegetation trend, through the remote sensing data, a paired analysis has been further conducted in the agricultural area adjacent to Turkey (Figure 7). A parallelogram-shaped region located in the Khabur River Basin is selected, whose vertex coordinates are [37.5°N, 40.5°E], [36.75°N, 41°E], [37.25°N, 39.5°E] and [36.5°N, 40°E] (a keyhole markup language (KML) file is attached in the supplementary data file). This region includes parts of Turkey and Syria, where irrigated farmland is widely distributed (Hole, 2009).

As shown in Figure 7, the correlation between EVI series of which the linear trends are removed is 0.8027 (p < 0.001), suggesting both areas are affected by similar, consistent climate variations. However, the vegetation exhibits contrasting trends of greening and browning in these two regions. The inter-annual variation trends of the EVI in Syria and Turkey sides in this basin are $-3 \times 10^{-3}$ and $4.5 \times 10^{-3}$, respectively. Similarly, the seasonal distributions of the variation trend also show significant differences. Therefore, it is likely that the effect of land management contributes to this difference. One possible reason is that with the increasing in agriculture input in Turkey, the land management has promoted the growth of farmland vegetation. The other reason maybe that, the social unrest in Syria has had a serious negative impact on agricultural production, and thus the land management has had a negative effect on the growth of farmland crops.
Recent vegetation changes have raised widespread concern. On one hand, vegetation is an important resource and ecological indicator. On the other, it also has an important feedback effect on climate, affecting the energy and water cycle of the earth-atmosphere system (Ryu et al., 2019; Xiao et al., 2019). Climatic driving factors for vegetation changes generally include temperature, precipitation and the fertilization effect of CO₂. Besides, the land-use change could also cause the change of vegetation. However, there is a serious lack of the studies on land management other than land-use changes (Prestele et al., 2017).

For the study area in this paper, due to the typical arid and semi-arid climate in this region, the primary constraint for vegetation growth is the water, therefore precipitation plays a dominant role here. However, the role of land management in this area is also worthy of attention. On the one hand, irrigated agriculture is widely distributed in this area (Figure 7(a)).
indicating high-intensity land management in this area. On the other hand, the economic fluctuation and social unrest in the
study area are relatively large. Among them, the most typical case is the fluctuation of the investment in land management
caused by social unrest, including irrigation facilities, seeds, fertilizers, pesticides, mechanization, effective field
management and so on (Mason, 2019; Mohammed et al., 2020).
Therefore, it is necessary to study the inter-annual fluctuation and long-term variation trend of vegetation in this area.
Because the study area is a typical water-limited ecosystem, the fluctuation of water conditions in this area will still have a
significant impact on farmland vegetation and natural vegetation even if there are irrigation facilities. For example, severe
droughts (Kelley et al., 2015) greatly affect the EVI during 2007–2009 (Figure 3). The average EVI of farmland vegetation
is 0.24 in 2001–2018, while it is 0.22 in 2007–2009. In contrast, the average EVI of natural vegetation is 0.15 in 2001–2018,
while it is 0.12 in 2007–2009. At the same time, droughts also lead to serious social impacts. Crop failure has triggered the
migration of people from rural to urban areas. In the suburbs of major cities in Syria, including Aleppo, Damascus, Deir ez-
Zor, Hama and Homs, there are more than 1.5 million people migrating from rural areas to cities and refugee camps, most of
whom are agricultural workers and family farmers (Eklund and Thompson, 2017; Gleick, 2014).
The long-term trend of vegetation, especially its spatial distribution, regardless of the residual analysis method or the rainfall
use efficiency, indicates that the vegetation decline in the Northern Syria is caused by human activities. This is naturally
linked to the devastating effects of the Syrian civil war in the region. According to the report of FAO (FAO, 2020), the
survey data of field research has revealed the comprehensive degradation of land management brought by war, including the
insufficiency of irrigation facilities, the lack of seed, fertilizers and pesticides, as well as the inadequacy of field management.
Therefore, several resources to demonstrated the map of social unrest during the Syria civil war since 2011. For instance, the
Armed Conflict Location and Event Data Project (ACLED) records the time and location information of unrest events,
including (not limiting to) the battle, pretests and violence (Raleigh et al., 2010). The Carter Center documents the territorial
changes of Syria at the community level since 2014 (Carter Center, 2020). Over a certainty region, the number of control
armed groups could also reflect the handover frequency and the degree of social unrest. These maps also have demonstrated
that the browning areas found here have experienced serious social unrest generally.

5. Conclusions
In this study, we investigated vegetation changes in Syria and their potential drivers. During 2001–2018, the EVIs of the
whole year and the growing period in the whole country decrease at the rates of $-0.8 \times 10^{-3}$ yr$^{-1}$ ($p < 0.1$) and $-1.9 \times 10^{-3}$ yr$^{-1}$
($p < 0.05$), respectively. Spatial patterns indicate that the EVI decline is mainly concentrated in the northern region. Both the
residual analysis and the RUE analysis suggest that besides climate change, human activities play a significant role in the
vegetation decline. A paired analysis in the Khabur River Basin across the border of Syria and Turkey demonstrates that both
sides of the region have widely-distributed croplands. EVIs on two sides have similar trend-removed inter-annual (climatic)
variations but contrary long-term trends, and the Syria side is browning while the Turkey side is greening. The spatial
patterns between the browning areas and social unrest of the civil var are generally coherent and relative reports have reported that social unrest had induced insufficient irrigation and lack of seeds, fertilizers, pesticides and field managements. These changes can be classified into the category of land management. Therefore, we concluded that the decline of vegetation condition (browning) in the north Syria is driven by the change of land management rather than climate change.

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