Using remote sensing to assess impacts of land management policies in the Ordos rangelands in China

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Abstract

Implementation of land management policies influences land use and hence causes environmental change. Taking the Ordos rangelands in China as a case study, this paper explores the potential of remote sensing to assess in dryland areas the impacts of policies on the environment. Thirteen Landsat images of the period 1978–2010 were acquired and those corresponding to the starting dates of implementation of different policies were selected for land cover change analysis; others were used to check the detected change and track the NDVI (Normalized Difference Vegetation Index) trajectory matched with time-series of meteorological data for calibration of natural response of rangelands to rainfall. The results indicate that policy impacts are complex and include both positive and negative aspects depending on the locality in space. On one hand, policies have aroused the enthusiasm of people in agricultural production and sand-control leading to recovery of about 2618 km$^2$ of desertified rangeland and sandy land, and economic growth; on the other hand, provoked vegetation degradation with an accumulated area of 2439 km$^2$ when policies cannot reconcile the conflict between environmental protection and the interest of rural people. However, degradation is not absolute and can be mitigated by implementation of rational policies.

Keywords: Land use/cover change, multi-temporal remote sensing, human intervention, spatial variability, NDVI trajectory

1. Introduction

In recent decades remote sensing has been widely applied in land cover change and land condition assessments, land degradation and desertification studies in the dry areas (Helldén 1988; Lambin and Strahler 1994; Diouf and Lammin 2001; Runnstrom 2003; Li et al. 2004; Wu 2003a, 2003b; Wessels et al. 2007; Hill et al. 2008; Zhou et al. 2008; Wu 2009; Del Barrio et al. 2010); and a number of publications have dealt with the linkages between environmental change and human activity (e.g. Geoghegan et al. 1998; Martens and Lammin 2000; Veldkamp et al. 2001; Verburg et al. 2002; Wu, Lammin, and Courel 2002; Lammin, Geist, and Lepers 2003; Geist and Lammin 2004; Aspinall 2004; Xie et al. 2005). Yet, few studies have focused specifically on the cause/effect relationships between the observed changes in land
use/cover and land management policies probably because of the intricate interlinkage between policies and the difficulty to distinguish and quantify the effect of one from others. In fact, the implementation of policies has been increasingly recognized as one of the major drivers of land surface change since they can directly or indirectly influence the land resources exploitation, creation of markets and economic opportunities. Responding to land use policies, the new socio-economic activities may lead to further environmental change and to new policy-making. Therefore, the analysis of policy impact is important and essential not only for the policy-makers but also for scientists. It allows them to better understand the role of policies in socio-ecological systems and to work out better strategies for sustainable land use and mitigation of land degradation in dryland ecosystems.

In the Ordos region, as in other provinces of China, different national and local land management policies have been implemented, especially in recent decades. This provides us with an opportunity to analyze the impacts of these policies on the environment in space and time. Although diagnostic studies of land cover change and land degradation have been undertaken in Ordos (e.g. Jia 1997; Gao, Zha and Ni 2001; Li et al. 2004; Runnstrom 2003; Wu, Zucca and Enne 2005), and Wu and De Pauw (2010) have attempted to reveal the linkage between policies and land degradation, there is an urgent need to have a complete view of policy impacts on the Ordos rangeland environment in order to provide useful reference for promulgation of more rational and sustainable land management policies in future.

Hence, the main objective of this case study in Ordos is to assess how remote sensing technology can assist land use planning by analyzing the impacts and implementation effectiveness of the past land management policies on the environment. This will be achieved by matching time series of remote sensing data with meteorological data, and socioeconomic information. Several remote sensing processing methods exist for analyzing changes in land use/land cover, such as change vector analysis, post-classification and differencing-and-thresholding techniques (Singh 1989; Lambin and Strahler 1994; Mas 1999; Lillesand and Kiefer 2000; Wu 2003b; Coppin et al. 2004; Wu 2007 and 2009). In order to be able to attribute the detected changes to human activity, temporary changes related to rainfall variations have to be taken into consideration. For this purpose, meteorological data have to be incorporated for calibration against the biophysical indicators such as NDVI (Rouse et al. 1973), as studied by Evans and Geerken (2004), and Wessels et al. (2007).

The most recent historical period, from the early 1950s to present, offers tremendous opportunities to study the effects of policy on the environment in Ordos. Since no remote sensing imagery is available for the period before the 1970s, our study is confined to the post-1973 period for which both satellite images and policy documents were available.
2. Materials and methods

2.1 The study area

2.1.1 Ordos Region

The Ordos Region is located in Inner Mongolia and is adjacent to the Shaanxi and Ningxia Provinces in China (Figure 1). Geomorphologically it is a part of the Ordos Plateau bordering the Loess Plateau on the south. The region is mainly covered with rangelands and situated in the transitional belt between the arid grasslands and deserts. The natural vegetation cover consists mainly of shrubs, perennial herbs and, very locally, forests (Wu, Zucca and Enne 2005). The soil types are entisols (sandy soils) in sandy land and deserts, mollisols in meadows and grasslands, gleysols in depressions or marshland, fluvisols in the alluvial plain of the Yellow River in the north, brown calcareous soils and loess in the hilly-and-gully area in the east. The average annual precipitation is around 317 mm. The eastern part of the region, e.g. Yulin, Ejihoro and Dongsheng, receives the highest annual rainfall with an average of about 335-409 mm. It gradually declines to the west and northwest, where the average annual rainfall is about 195-224 mm in Yinchuan, Shizuishan and Linhe (see location in Figure 1). With a summer rainfall pattern, 70% of the annual rainfall is concentrated in the period July-September. Winds from the northwest blow on average 230 days per year and may exceed Beaufort Scale Force-8 (speed >17 m s\(^{-1}\)) during more than 40 days. The concentrated rainfall and strong wind provoke soil erosion and water loss (Xu 1999, 2006). The study area, covering a territory of 17,000 km\(^2\), is located in the southwest part of the Ordos Region, and more specifically, an area including a part of the two Banners namely Otog and Otog Front, and Wuhai City in Inner Mongolia, and Taole County in Ningxia (Figure 1).

In terms of natural resource endowment, the Ordos Region has abundant good-quality coal resources as well as natural gas and oil reserves. In recent years Ordos has become one of the national energy bases under the Chinese Government’s mid-to-long term national strategy “To Develop the West”. Significant land use/cover change and land degradation have taken place in Ordos owing to long-time human activity in grazing, deforestation and land reclamation for agriculture driven by a number of local and national policies (Jiang et al. 1995; Jia 1997; Chen and Gao 2001; Zhang and Wang 2001; Wu 2003a; Xu 2006).

Figure 1. Location of the study area and distribution of the meteorological stations.

The basic administrative unit shown in this figure in Inner Mongolia is Banner, which is similar to the unit “County” in other provinces. The location of meteorological stations are marked by a circle.
2.1.2 Land reclamation and degradation cycles through history

According to the studies of Jiang et al. (1995), Chen and Gao (2001), and Zhang and Wang (2001), cultivation in Ordos dates back to the late Zhou Dynasty (474-221 B.C.). Policy-driven land reclamation started in the Han Dynasty (206 B.C.-220 A.D.), when Emperor Wu issued an order in 120 B.C. “Migration for consolidating the border areas”. Hence, about 700,000 peasants emigrated from the interior of the country to the Ordos and Loess Plateaus for land reclamation, and at the same time consolidating the border areas (Enkhee 2003). Widespread forest and grasslands were hence reclaimed for agriculture. In the Tang Dynasty (618-907A.D.), similar policy was implemented in the period 707-820 A.D., leading to forest-clearing for construction of the frontier fortresses and land reclamation for crop production which gradually resulted in land degradation — a series of small incursions of sandy land — the initial stages of the Hobq Desert and Mu Us Sandy Land (Enkhee 2003). In the Qing Dynasty (1644-1911 A.D.), again under the similar policy, deforestation for construction of the temples and government houses and land reclamation for cropland took place (Zhang and Wang 2001). In recent decades, three peaks of land reclamation from grassland to cropland occurred again in 1955-1956, 1958-1962 and 1971-1973 under the new policies such as “Giving the prominence to food production” (Chen and Gao 2001). Owing to the low productivity of these converted rangelands (annual rainfall < 250 mm), the fields were abandoned after two to three years’ cultivation and became degraded by topsoil loss due to exposure to the strong winds and water erosion by concentrated rainfall. This loss of land necessitated new reclamation in the grasslands elsewhere.

After 1973, several periods can be differentiated with respect to how national and local government land policies have influenced land use in the Ordos rangelands:

Period 1 (1979-1985):

This period refers to the nationwide implementation of Deng Xiaoping’s “Household land tenure policy” in 1979-1984, and the promulgation of the first law concerning grassland use, the “Decree of grasslands”, in 1985. The main objective of the first policy was to stimulate people towards agricultural production in order to minimize the risk of famine in the country after the “Cultural Revolution” (1966-1976). Deng’s land tenure policy was fully implemented in all agricultural provinces, and food production was much increased in the whole country. In Ordos, an agropastoral region, this policy transferred the tenure of agricultural land and ownership of production to individual farmers, resulting in new land reclamation and boosts of food production. For example, in the Banner Otog Front, food production increased by 163% while population grew only by 15.5% from 1980 to 1989 (Statistic Yearbook of Otog
Front, 2009). On the other hand, the Grasslands Decree of 1985 kept the grasslands as a collective resource whereas livestock was private (Erdenzhab 2002; Gai 2007).

Period 2 (1987-1992):

After the countrywide success of the agricultural reform, Deng further proposed a national “Open and reform policy” and issued the “Decree on legalization of the private economy” under the strategy named “Invigorating the domestic economy and opening to the outside world” in 1987-1988 (In memory of Deng Xiaoping: http://cpc.people.com.cn/GB/69112/69113/69683/index.html). However, nationwide implementation of these policies was not conducted until February 1992 when Deng paid an official visit to the southern regions of the country. These governmental actions promoted the privatization of the collective or national enterprises and encouraged the development of the private economy. In response to these new policies, countless local private companies were set up in Ordos after 1992 and started to bring into the national economy local products such as coal, oil, gas, kaoline, as well as wool, and medicinal and edible wild herbs (Ephedra sinica, Glycyrrhiza uralensis, Nostoc commune var. flagelliforme). At the same time self-organized teams and family-based companies, composed of a great number of peasants and herdsmen, started to invest in the control of land degradation and to increase the economic value of the land, with the encouragement of existing governmental agencies for combating desertification. As expected, the economy in Ordos developed rapidly in the 1990s. Taking the same example of the Banner Otog Front, the Gross Domestic Product (GDP) from 1989 to 1999 increased by 614%, and the annual income of rural people augmented by 628%. The drawback of these two policies was their exclusive focus on pursuing economic development and profit, and neglect of environmental protection. As a result, overexploitation of land resources (e.g. coal mining and oil exploitation without protection measure, overgrazing, excessive collection of medicinal and edible herbs) lead to widespread degradation of the fragile vegetation cover.

Period 3 (2000-2001):

Aware that as the primary vocation for Ordos, the region had become to serve as one of the national energy bases due to its abundant fossil fuel and gas resources, and that earlier agricultural land use policies had led to severe land degradation, the Chinese Government issued new legislation for the area. The Central Government promulgated in 2000 a decree “Prohibiting the collection of the medicinal and edible herbs (e.g. licorice and ephedra) in grasslands”. In 2001, on top of this national order, the local governments of Otog and Otog Front published “Grazing ban and rotation policy with subsidy system” as well as the call “To return cultivated land to grassland”. Government subsidies encouraged herdsmen to utilize, in the areas affected by the grazing ban, some highly productive land to cultivate aridity- and cold-resistant forage grasses, such as alfalfa (Medicago sativa) and Astragalus adsurgens pall for breeding
animals and preserving large surfaces of grassland. The previously open grazing system was then converted into a system based on indoor dry-lot feeding. Other areas were designated for rotational grazing.

2.2 Data

Multi-temporal Landsat MSS, TM, ETM+ images from 1978 to 2010 were obtained (Table 1). The images of 1978, 1989 and 1999 were considered as the initial states of different policies implementation periods. To represent the current state the image of 2007 was used since the 2009 ETM+ image contains a defect due to the Scan Line Corrector (SLC-Off) problem, and there is considerable cloud cover (>33%) in the TM 2010 image. However, both images could still be used with others for projecting the NDVI trajectory of the test sites.

Annual rainfall data for the years 1991, 1996, 1999 and recent years 2005-2009 were collected for 16 stations within the Ordos Region, especially, in the Banners of Otog and Otog Front and Yinchuan Region in Ningxia from the Statistic Yearbooks of Ningxia and of the Banner Otog Front. Monthly rainfall data series of Otog and eight adjacent stations (Dongsheng, Linhe, Ejinhoro, Yanchi, Yinchuan, Yongning, Yan’an and Yulin) from 1960s to 2010 were acquired from the China Meteorological Data Sharing Service (see Figure 1 for location).

Together with the acquisition dates of the different satellite images, the implementation onset dates of different land use policies are shown in Table 1.

Table 1. Multi-temporal satellite images used in this study

2.3 Approaches to discern land surface change

Generally, classification-related approaches are most frequently applied to assessing land cover change. However, as many authors have argued, these approaches may produce huge errors unless each class is accurately classified and validated (Lambin and Strahler 1994). With the advances in classification techniques, a number of successful studies have been reported (Mas 1999; Lillesand and Kiefer 2000; Wu 2009). Another practical approach is the differencing-and-thresholding techniques (Jensen and Toll 1982; Lambin and Strahler 1994; Lillesand and Kiefer 2000; Wu 2003b; Coppin et al. 2004; Wu 2009), which can identify accurate changes in space and time. Therefore, these two methods have been jointly employed for land cover change detection in this study. The processing steps include:

2.3.1 Preprocessing
The Landsat ETM+ image dated August 12, 1999 was geometrically corrected using topographic maps on the scale of 1/200,000 and GPS points obtained during the field investigation in August 2000 and June 2002 (for a total of 121 ground control points) in the WGS84 datum and projection UTM Zone 48N using the 3rd order polynomial model and nearest neighborhood resampling. This image was used as reference to correct the other images with RMS error 0.23-0.58 pixels. Some recent images (2002, 2006, 2007, 2009 and 2010) obtained from USGS are already well georeferenced. The correction procedure was then avoided.

An image-based approach, the COST-model proposed by Chavez (1996), was introduced in this study to correct the additive and multiplicative atmospheric effects in the Landsat MSS, TM and ETM+ images; the dark-object subtract (DOS) method (Chavez 1988) was used to determine the haze in each band.

The most frequently applied vegetation index, NDVI, was derived for land cover change detection and trajectory analysis. It is expressed as:

\[
\text{NDVI} = \frac{\rho_{\text{NIR}} - \rho_R}{\rho_{\text{NIR}} + \rho_R} \quad (1)
\]

where \(\rho_{\text{NIR}}\) and \(\rho_R\) are respectively the reflectance of the near infrared and red bands of TM, ETM+ and MSS images.

The MSS image dated 1978 was resampled into 30 m resolution using nearest neighborhood resampling approach to have more comparability with others. Then two MSS NDVIs from two different near infrared (NIR) bands (band 3: 0.70-0.80 \(\mu\)m and band 4: 0.80-1.1 \(\mu\)m) vs. the same red band (band 2: 0.60-0.70 \(\mu\)m) were produced and averaged to get the mean NDVI, which is more compatible with the NDVI of TM/ETM+ images.

2.3.2 Indicator differencing and thresholding

To distinguish a negative trend of vegetative greenness from a positive trend, a differencing was applied to the atmospherically corrected NDVI images between the periods 1978-1989, 1989-1999 and 1999-2007 followed by a thresholding technique. More specifically, the differencing procedure consists of subtracting the NDVI of the two different dates. Taking the period 1999-2007 for example, the differencing can be expressed as:

\[
\Delta \text{NDVI} = \text{NDVI}_{2007} - \text{NDVI}_{1999} \quad (2)
\]

The range of \(\Delta \text{NDVI}\) is between -2 and 2. After statistics, we got the mean (M = 0.056), standard deviation (\(\sigma = 0.137\)), minimum (Min = -1.997), and maximum (Max = 1.963) values of the differenced image, of which the histogram shows a leptokurtic distribution. The pixels between M-\(\sigma\) and M+\(\sigma\) take up 82.76% of the whole image, indicating that the majority of pixels have no change. The thresholds (T) for positive change (greenness increase) and negative change (greenness decrease) from 1999 to 2007 can
be determined by testing the potential thresholds respectively in the two tails of the ΔNDVI histogram. To get the positive change threshold, a tuning was conducted while moving outward from M+σ to the positive end, by adding a (1/8)σ each time; and similarly, we can obtain the negative change threshold while testing it from M-σ to the negative end by reducing a (1/8)σ each time. Each tentative threshold was checked against the color composites of 1999 and 2007 for all pixels till when the most relevant thresholds were obtained.

The same change detection technique was applied on the ΔNDVI of the periods 1989-1999 and 1978-1989. Some changes due to drying up of water bodies (NDVI also increases for change from water to land) were not real vegetation cover change; other changes related to agricultural rotations such as conversion from cropland to fallow land or vice versa were not the subject of this analysis. All these false change were excluded manually by mask operation.

The advantage of this change detection technique lies in the fact that it can exactly position the changes in space and reveal the variation in greenness extent, and can thus further assess the dynamics of biomass production. For this reason, the conversion of grasslands into croplands and pastures as well as grassland in rotation (fallow), can also be identified in each observation period. The disadvantage is its insensitivity to the change in very low greenness land cover where there is no significant difference in NDVI. For example, recent sand-control with very low greenness cannot be identified out in the deserts.

2.3.3 Post-classification differencing

To avoid the underestimation, another change detection technique “post-classification differencing” (Wu 2009) was applied to the images of MSS 1978, TM 1989, ETM+ 1999 and TM 2007 as a complementary to the differencing-and-thresholding approach. Prior to the differencing step, a supervised classification using the Maximum Likelihood classifier was carried out in which the training areas were visually selected in the color composites using Google Earth as reference. Four major land cover types were differentiated: sands (deserts), vegetation cover (VGT, including agricultural patches, pastures and grasslands with low greenness), water bodies and bare soils (including normal bare land, salty crust or saline land occurring near rivers, lakes and swamps). These classes were further aggregated into two major groups: sands and non-sands (VGT cover + water bodies + bare soils) after classification. The overall accuracy against the ground truth “regions of interest” and Kappa Coefficient of each individual classification before aggregation are shown in Table 2.

Table 2. Overall accuracy of each individual classification
Differencing on these classes at different times allowed us to assess whether a given class expanded or contracted during the study period. For example, where Sand1989 – Sand1978 > 0, it was considered that the sand class had extended or previous grassland (or other non-sand areas) was covered with sands (deserts) for the given area, whereas a negative difference indicated that it had contracted or former desert was covered with vegetation (e.g. due to sand-control).

The advantage of this technique is that it can detect both subtle changes such as sand-control, desert extension as well as abrupt change (e.g. conversion from grassland into urban or croplands with significant difference in greenness). The disadvantage is that it cannot precisely quantify the change degree of the greenness.

The two approaches “differencing-and-thresholding” and “post-classification differencing” are complementary and their combined use leads to optimal differentiation of land cover change and land degradation.

2.3.4 Identification of specific land cover change types

To understand the spatial variation in land use/cover changes, density slicing on the differenced NDVIs and masking operation to the detected changes were applied to identify, against the original image composites, the specific change types in different periods such as extension of cropland and pasture, extension of coal mining area, sand-control, and vegetation degradation, etc.

2.4 Calibration with meteorological data

To calibrate the detected change, we have to first understand the natural response of ecosystem to rainfall, and then to distinguish the role of human intervention. Malo and Nicholson (1990), Tucker, Dregne, and Newcomb (1991), Davenport and Nicholson (1993), Wang et al. (2001), Wessel et al. (2007), Udelhoven et al. (2009) analyzed the vegetation response to rainfall and concluded that monthly maximum NDVI is best correlated with the rainfall in the concurrent plus two antecedent months in Africa, USA and Spain; Celis, De Pauw and Geerken (2007) found that in the Fertile Crescent an excellent agreement between the average NDVI and a linear combination of the average annual rainfall and the minimum temperature of the coldest month; and Evans and Geerken (2004) identified in Syria that the rainfall period is best correlated to the annual maximum NDVI (NDVI_{max}) or peak NDVI (NDVI_{peak}) which represents the total annual green biomass production.

In the light of the above knowledge, the following points were taken into consideration in our calibration:
Landsat images used for this land cover change detection were acquired in August, corresponding to the major annual rangeland biomass production and rainfall period. Thus the reflectance-based NDVI derived from these images is considered as the annual NDVI$_{\text{max}}$, which is representative of the total green biomass (Evans and Geerken 2004);

The accumulated rainfall in the biomass production season (May to September) in the study area accounts for 85.3% of the total annual rainfall in the observed years. Both the growing season rainfall and annual rainfall can be used to calibrate the response of the NDVI$_{\text{max}}$. We used the annual rainfall for this purpose due to the lack of monthly data from some stations;

The least-square regression model is applicable in such kind of calibration. Instead of using the residuals trends (Evans and Geerken 2004; Wessels et al. 2007) as human-induced indicator, we consider the correlation coefficient as a direct indicator. As argued before, high correlation between NDVI$_{\text{max}}$ and annual rainfall implies the natural response of vegetation to rainfall with little or without disturbance; however, low correlation is considered as a result of human intervention such as overgrazing or other kinds of land modification overlapping on the rainfall. In both cases, inter-annual variability of rainfall is already considered.

The calibration was conducted in two ways shown in the following sections.

2.4.1 Diachronous calibration

To analyze the biophysical response of ecosystem to rainfall and human intervention over time, six sites (A, B, C, D, E and F) were selected on the basis of whether evident land cover change had occurred or not during the three observation periods. Their location and features are shown in Table 3 and Figure 3.

### Table 3. Features of the selected sites

For each site the averaged reflectance-based NDVI value of 13 years was respectively computed to compose the NDVI trajectories (a, b, c, d, e and f in Figure 2). The annual rainfall data were spatially interpolated using the Inverse Distance Weighting (IDW) approach to produce the annual rainfall raster layers from which the average rainfalls of each site were extracted for the observed years. The average NDVI values of each site were then plotted against the corresponding annual rainfall data (Figure 2) and a least-square regression analysis at the confidence interval of 95% was undertaken to reveal the relationship between NDVI trend and rainfall in these sites (a’, b’, c’, d’, e’ and f’ in Figure 2).

2.4.2 Cross-spatial analysis
This analysis was aimed to ascertain the spatial variation of the rangeland ecosystem in response to rainfall by relating the vegetation greenness (e.g. NDVI) to the rainfall distribution in space at one point of time (one year in this case). The NDVI images and their corresponding interpolated rainfall data (raster layer) of 1978, 1989, 1999 and 2007 were chosen for this purpose. In order to have spatial representativeness and quick assessment, 2000 points were randomly generated in the whole study area to cover all types of land cover. After excluding the points falling in the irrigated agricultural land and desert patches (in which the NDVI has theoretically no relationship with rainfall), 1485 points remained to which a linear least-square regression model was fitted.

**Figure 2. NDVI trajectories and annual rainfall**

Note: (a) and (a') for Site A; (b) and (b') for Site B; (c) and (c') for Site C; (d) and (d') for site D; (e) and (e') for Site E, and (f) and (f') for Site F

### 3. Results and discussion

Through the above processing and analysis, the results obtained are presented here for discussion.

**Diachronous calibration at test sites**

The NDVI trajectory at site A (Figure 2a) shows a decrease in the period 1978-1989, followed by an increase up to 2002. From the texture in the ETM+ images dated 1999 and 2002, it appears that site A and its surrounding area were controlled and managed zones from 1996 to 2002: the “network belts” in which local people planted grasses and shrubs to combat desertification could be clearly discerned in both images. From 2004 to 2007 the NDVI value decreased again whereas annual precipitation increased. This “unnatural” NDVI drop is most likely the result of human intervention, e.g. overgrazing.

Site B was selected in the natural grassland, protected and with little disturbance by human activity up to 2004, but opened for grazing after 2005/2006. Although the overall correlation coefficient between the NDVI trajectory and rainfall is low ($R^2 = 0.405$, Figure 2b'), if we consider only the observed years before 2004, the fluctuation in NDVI is well consistent with that of annual rainfall and the coefficient is rather high ($R^2 = 0.592$).

Site C, protected from grazing in the period 1996-2006, presents a similar pattern as Site B (Figure 2c). With the opening for grazing in 2007, NDVI declines, but no significant degradation was detected in this site.
Site D experienced degradation in the period 1989-1999, particularly in the years from 1996 to 1999. The NDVI trajectory also fluctuates with rainfall but the correlation is low ($R^2 = 0.264$). The recovery of the grasslands occurred in the period 2002-2007.

Site E was carefully selected grasslands where neither evident degradation nor conversion to pastures was observed. Hence the NDVI trajectory and annual rainfall in site E shows the best correlation ($R^2 = 0.664$) among the all test sites.

Site F is similar to Site E, but some land management, e.g. conversion from grasslands into pastures in 2004 and low greenness in comparison with the rainfall in 2010 were discerned. For this reason the correlation between the NDVI and rainfall is lower than that of Site E.

In summary, the diachronous calibration reveals that the correlation between NDVI and annual rainfall is generally weak. The $R^2$ values vary from 0.044 (no correlation) to 0.664 (moderately high correlation), depending on the extent of human intervention (grazing, overgrazing, conversion of rangeland into cropland, etc.). In Figure 2a, 2b, 2c and 2d, the “unnatural” points represent the occurrence of land use/cover change, an intervention of human activity.

**Cross-spatial analysis**

The linear regression analysis did not show any correlation between the NDVI and rainfall distributions at 95% confidence level in each observation year ($R^2 = 0.002 - 0.111$). This implies that in the study area the vegetation greenness, represented by NDVI, an important spatial indicator of rangeland condition (Del Barrio et al. 2010), cannot be explained by the distribution of rainfall only, and is therefore more influenced by land use practices, superimposed on the rainfall pattern.

**Impacts of land use and rainfall on grasslands**

One important result obtained through the above two calibration analyses is that the peak greenness of rangeland is a combined effect of both rainfall and human activity — land use. Without human intervention, the greenness expressed by the maximum NDVI, is positively correlated with the annual precipitation, in our case, $R^2 = 0.664$. However, the increased human intervention led to a lower correlation between the two factors. Hence, the NDVI-Rainfall models provide a simple but effective view to assess the impacts of land use and management on rangelands.

**Land cover changes**

The relevant land cover changes and their change rates corresponding to the different policy implementation periods are shown in Figures 3. In general the rangelands in the study area have
experienced two kinds of change in each period: one is vegetation greenness growth (VGT-G) and the other is vegetation degradation (VGT-D) as shown in Figures 3.

**Figure 3. Rangeland biophysical changes in the study area in the period 1978-2007**

Note: The figure shows (1) vegetation vigor growth (VGT-G), and (2) vegetation degradation (VGT-D). Sites A, B, C, D, E and F were selected to calibrate the relationship between biophysical feature changes (e.g. NDVI) and rainfall variation.

VGT-G, with a total surface area of about 2618 km\(^2\) or 15 % of the total land area in the period 1978-2007, is attributed to different types of conversion of the grasslands into more productive land cover categories, such as: (1) croplands, pastures/forage land (e.g. areas planted with alfalfa) and cash crop plantation (planted with ephedra and licorice), with a total area of 557 km\(^2\) in the period 1978-2007 and irrigated by either the Yellow River (e.g. in Taole and Wuhai) and its tributary or by groundwater (e.g. in the communes of Arbas and Shanghaimiao); (2) sand-control by soil conservation methods, including afforestation (e.g. planting poplar, pine and *pinus sylvestris var. mongolica*) and cultivation of drought-resistant shrubs and herbs (such as *Hedysarum leave*, *Caragana korshinski*, *Artemisia sphaerocephala* and *Artemisia ordosica*, sea-buckthorn) to combat desertification, improve land condition, and generate economic value; and (3) grasslands in fallow (in rotation without grazing) leading to a recovery of the rangelands.

VGT-D as shown in Figure 3 is characterized by a decrease in vegetation greenness in the rangelands due to: (1) overgrazing in the permitted rotation grasslands, especially during the periods of droughts (the major degradation type); (2) overgrazing and trampling of animals around settlements and water points (secondary degradation type); (3) desert dynamics as expressed by mobile sand dunes in the western part of the study area (north of Taole, northwest of Shanghaimiao, southwest of Arbas), moving along the wind-blowing direction mainly from northwest to southeast at an annual rate of about 11-21 m depending on locations. In the southern parts of the study area this desert encroachment has been reduced or even stopped as a result of sand-control in the past decades; and (4) coal mining in Wuhai and Otog (Qipanjin) in the recent decade. These kinds of land degradation have caused the temporary loss of vegetation cover in 2439 km\(^2\) of rangelands (13 % of the total land) in the period 1978-2007.

The change rates for each identified specific change in each policy implementation period are summarized in Table 4 and Figure 4; the spatial extent of each specific change category is shown in Figure 5.

| Table 4. Land cover changes and their change rates in the study area |
It was also noted that grassland degradation did not occur in the same place all the time but moved from one place to another in different periods as a result of the traditional grazing habit, and more recently linked to the implementation of the “Grazing ban and rotation policy”. In most cases after degradation recovery occurred.

**Discussion**

As revealed in the above sections, about 28% of the total territory of the study area has experienced change corresponding to the implementation of different policies in the past three decades. A brief discussion on the interaction between the change in environment and policies is given below.

After the first period of new policy formulation (1979-1985), satellite images provide evidence that in the period 1978-1989 land reclamation had taken place at a rate of 7.45 km² y⁻¹ (see Table 4, Figure 4 and 5a), which coincided with a dramatic increase in food production (for example, 163% in the Banner Otog Front). The predominant biophysical change in this period is vegetation degradation, as a result of grazing activity in the collective grasslands in the western parts of the study area authorized by the Grasslands Decree in 1985 (see Figures 4 and 5e, and Table 4). Another factor leading to degradation is the fuel wood collection in the desert rangelands and mountainous slopes. It was reported that before the start of large-scale coal mining in Ordos in 1990s, collection of shrubs (*Artemisia ordosica*) as fuel wood for cooking and heating was widespread (Chen and Gao 2001; Wu 2003b). The global degradation rate during this period was estimated at about 60 km² y⁻¹ (see Table 4).

During the 2nd policy implementation period in 1990s, the local outcome of the burst in economic activity was mixed. Landsat images, dated 1989 and 1999, show an increase in NDVI (1989-1999 VGT-G in Figure 3), and indicate that some areas of degraded land and deserts in the southwest were controlled and that a considerable part was converted into croplands and cash crop plantations such as ephedra and licorice, especially in the Shanghaimiao and Arbas Communes (Figures 4 and 5a, 5b, 5c). In the Banner Otog Front alone, food production was increased by 448.7% in the period from 1989 to 1999, which
corroborates the evidence from remote sensing of the extension of cropland (see Table 4). On the other hand, the incentives to improve family income and livelihoods and business opportunities offered by the new policy “Invigorating the domestic economy”, resulted in the widespread collection by local people of wild medicinal and edible herbs. This factor, combined with overgrazing and fuel wood collection during a period of rainfall reduction (e.g. 1999), can offer an explanation for the significant vegetation degradation observed by remote sensing (Figures 3, 4 and 5e). Thus the immense increase of the net income of the rural people by 628% and of the GDP by 614% from 1989 to 1999 in the Banner of Otog Front was to a large extent at the cost of grassland degradation, which was expanding at a rate of about 143 km$^2$ y$^{-1}$. In the meantime, coal mining by national and private companies started to be accelerated mainly in Wuhai and Qipanjing in the Banner of Otog (Table 4 and Figure 5d).

Since 1999, thanks to the decree “Prohibiting the collection of the medicinal and edible herbs in grasslands” and the “Grazing ban and rotation policy with subsidy system”, the greenness in land cover was significantly increased in the whole study area (1999-2007 VGT-G in Figure 3, and Figures 5a, 5b, and 5c). The specific land management measures taken to combat degradation included rangeland recovery, pasture extension and sand control, which were implemented at the rates of about 110 km$^2$ y$^{-1}$, 33 km$^2$ y$^{-1}$ and 52 km$^2$ y$^{-1}$ respectively. In comparison with the previous period 1989-1999, degradation has been significantly reduced, and its rate of growth was only about 14 km$^2$ y$^{-1}$. Meanwhile, from 1999 to 2009, the cattle numbers and food production did respectively increase in the two Banners Otog and Otog Front by around 51% and 106%, resulting in an increase of herdsmen’s incomes by 267%.

In summary, remote sensing reveals that land use and management policies exerted both positive and negative impacts on the rangelands in the West Ordos. The policies implemented in the three observed periods have clearly awakened the enthusiasm of peasants, improved the agricultural production, promoted the emergence of hundreds of private enterprises including the “sand-control teams”, lifted the regional economy as well as household incomes, and ameliorated the local environment. However, degradation could not be stopped when policies (e.g. those in 1985 and 1987-1988) were only aimed at keeping grassland tenure collective without considering the responsibility of herdsmen, or only at economic development without environmental protection measure or at environmental preservation without any benefit for rural people. Degradation just moved from one place to another following the loopholes in the legislative environment or weaknesses in the enforcement mechanisms. When the implementation of policy, even an enlightened one, could not be well monitored or enforced, or when it conflicted with the personal interests of local people (Erdenzhab 2002; Gai 2007), land degradation is inevitable since herdsmen would be more interested in increasing short-term benefit rather than protecting the rangelands from degradation.
4. Conclusions

This paper presents a cross-cutting study to assess the impacts of land use and management policies on the rangelands in Ordos by analyzing satellite images at times that correspond to the initial stages of the implementation of different policies. It was noted that in the study area land degradation, although it could be favored by natural factors such as drought and wind blowing, was basically the outcome of policies that did not take into consideration the fragility of the ecosystem, such as the land reclamation works in the early of 1970s (Zhang and Wang 2001; Enkhee 2003). Some activities under a rational but poorly implemented policy also caused local land degradation, e.g. overgrazing in the permitted areas under the “Grazing ban and rotation policy”. At the same time the history of the study area demonstrates that where sound policies and effective implementation go hand in hand, mitigation of land degradation and rangeland recovery are feasible. The legalization of the Chinese private economy in 1988 made the sand-control movement by individual families in the region possible and this in turn contributed to the growth in economy and significantly to the restoration of the sandy land and desert ecosystems. The grazing ban policy coupled to a subsidy system did restore and protect a large part of the grasslands and at the same time improve the income and livelihood of the rural people. Therefore, the impacts of policies are complex, depending on the extent, depth and duration of their implementation, but also on their ability to reconcile the desire for a healthy agropastoral ecosystem with the need to improve the livelihoods of poor people who depend on this resource base. If there is a contradiction, land degradation is unavoidable.

In conclusion, although of a reconnaissance nature, this study shows the great potential of remote sensing to evaluate the complex impacts of policy implementation on the rangeland ecosystem by monitoring the changes in biophysical features and land cover, and linking this information to the social and political environment. Interpreting the impacts of a single policy is difficult and very challenging, particularly, when its implementation overlaps with other policies promulgated in the same period. What can be revealed is the global effect of the major one or the joint effect of several ones. The proposed remote sensing methodology, consisting of an integrated processing procedure followed by a calibration with meteorological data, produced relevant results that revealed well the joint effect of policies in each period. Future research shall go in the direction of a more comprehensive and quantitative assessment with more reasonably integrated sets of assessment indicators (Sommer et al. 2011; Zucca et al. 2012), through a spatial analysis on the extent, depth and duration of the policy implementation linked to biomass and socio-economic data at commune- or village-level.
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Table 1. Multi-temporal satellite images used in this study

| Captors          | Scene frame | Acquisition dates | Spatial resolution | Policy implementation period |
|------------------|-------------|-------------------|--------------------|------------------------------|
| Landsat 5 TM     |             | August 18, 2010   | 30 m               |                              |
| Landsat 7 ETM+   |             | August 07, 2009   | 30 m               |                              |
| Landsat 5 TM     |             | August 10, 2007   | 30 m               | Current state                |
| Landsat 5 TM     |             | August 07, 2006   | 30 m               |                              |
| Landsat 5 TM     |             | August 17, 2004   | 30 m               |                              |
| Landsat 7 ETM+   | Path-Row: 129-33 | August 20, 2002     | 30 m               |                              |
| Landsat 7 ETM+   |             | August 30, 2000   | 30 m               |                              |
| Landsat 7 ETM+   |             | August 12, 1999   | 30 m               | Initial state of Period 3    |
| Landsat 5 TM     |             | August 11, 1996   | 30 m               |                              |
| Landsat 5 TM     |             | August 30, 1991   | 30 m               |                              |
| Landsat 5 TM     |             | August 24, 1989   | 30 m               | Initial state of Period 2    |
| Landsat 5 TM     |             | September 20, 1987| 30 m               |                              |
| Landsat 3 MSS    |             | August 21, 1978   | 56 m               | Initial state of Period 1    |

Table 2. Overall accuracy of each individual classification

| Image acquisition date | Overall accuracy (%) | Kappa Coefficient |
|------------------------|-----------------------|-------------------|
| August 21, 1987        | 99.65                 | 0.988             |
| August 24, 1989        | 98.70                 | 0.967             |
| August 12, 1999        | 97.39                 | 0.918             |
| August 10, 2007        | 98.91                 | 0.966             |
Table 3
Features of the selected sites

| Site | Longitude   | Latitude    | Surface area (ha) | Major features                                                                 |
|------|-------------|-------------|-------------------|-------------------------------------------------------------------------------|
| A    | 106.6928 E  | 38.4936 N   | 2317              | Degradation in 1978-1989, recovery due to rotation and sand-control in 1996-2002 and degradation again in 2004-2007 |
| B    | 107.3907 E  | 38.3194 N   | 1094              | Protected natural rangeland without disturbance of grazing activity before 2004, but open for grazing since 2005/2006 |
| C    | 107.3885 E  | 38.6881 N   | 1094              | Managed grassland area but open for grazing after 2007                          |
| D    | 107.6824 E  | 38.9754 N   | 1301              | Degradation in 1989-1999, especially in 1996-1999 but recovery after 2002       |
| E    | 108.2800 E  | 38.68857 N  | 480               | Natural rangeland without significant change and little influenced by grazing in the past 32 years |
| F    | 107.637 E   | 38.61615 N  | 374               | Similar to site E, mostly unmanaged but more or less influenced by grazing and conversion of some grassland patches into pastures after 2004 |
Table 4
Land cover changes in the study area

| Policy implementation periods | VGT-G | VGT-D |
|------------------------------|-------|-------|
|                              | Crop & pasture extension | Sand control | Rangeland recovery | Desert extension | Rangeland degradation | Coal mining |
| Area (km²) | Rate (km² y⁻¹) | Area (km²) | Rate (km² y⁻¹) | Area (km²) | Rate (km² y⁻¹) | Area (km²) | Rate (km² y⁻¹) | Area (km²) | Rate (km² y⁻¹) |
| 1978-1989 | 81.97 | 7.45 | 82.48 | 7.50 | 153.28 | 13.93 | 80.66 | 7.33 | 665.04 | 60.46 | 5.93 | 0.54 |
| 1989-1999 | 215.15 | 21.52 | 97.61 | 9.76 | 429.26 | 42.93 | 78.1 | 7.81 | 1429.36 | 142.94 | 9.54 | 0.95 |
| 1999-2007 | 260.57 | 32.57 | 418.78 | 52.35 | 878.78 | 109.85 | 38.28 | 4.79 | 114.00 | 14.25 | 18.28 | 2.29 |

Note: y represents “year”.
Figure 1. Location of the study area and distribution of the meteorological stations.

The basic administrative unit shown in this figure in Inner Mongolia is Banner, which is similar to the unit “County” in other provinces. The meteorological stations are marked by a circle.
Figure 2. NDVI trajectory and annual rainfall

Note: (a) and (a’), (b) and (b’), (c) and (c’), (d) and (d’), (e) and (e’), (f) and (f’) for Site A, Site B, Site C, Site D, Site E, and Site F, respectively.
Figure 3. Rangeland biophysical changes in the study area in the period 1978-2007

Note: The figure shows (1) vegetation vigor growth (VGT-G), and (2) vegetation degradation (VGT-D). Sites A, B, C, D, E and F were selected to calibrate the relationship between biophysical feature changes (e.g. NDVI) and rainfall variation.
Figure 4. Land cover change rates in the observed policy implementation periods
Figure 5. Specific land cover changes in the study area