Geographical Information Based Evaluation System for Drought

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Abstract: Drought is one of the major natural hazards affecting the environment and economy of countries worldwide. As a natural hazard, drought is best characterized by multiple climatological and hydrological parameters. An understanding of the relationships between these two sets of parameters is necessary to develop measures for mitigating the impacts of droughts. Droughts are recognized as an environmental disaster and have attracted the attention of environmentalists, ecologists, hydrologists, meteorologists, geologists, and agricultural scientists. Low rainfall has mainly caused droughts and subsequent reduction in agricultural production. Droughts have been a recurring feature of the Indian climate, therefore, study of historical droughts may help in the delineation of major areas facing drought risk and thereby, management plans can be formulated by the government authorities to cope with the disastrous effects of this hazard. In recent years, Geographic Information System (GIS) and Remote Sensing (RS) have played a key role in studying different types of hazards, either natural or man-made. This paper emphasizes upon the application of RS and GIS in the field of drought risk evaluation. The study area taken is a part of the Jamnagar district of Gujarat between latitude 22°19′46″N to 22°46′01″N and longitude 70°20′56″E to 70°47′34″E. The study was conducted with satellite images of year 1977, 1990 and 1999. Data has been acquired mainly from two sources, firstly Normalized Difference Vegetation Index (NDVI) obtained from satellite sources and secondly rainfall obtained from ground rainfall stations record. In the present paper, an effort has been made to derive drought risk areas facing agricultural as well as meteorological drought by use of Normalized Difference Vegetation Index (NDVI) from Landsat images. NDVI values reflect the different geographical conditions quite well. The NDVI and rainfall was found to be highly correlated. It is therefore concluded that temporal variations of NDVI are closely linked with precipitation.

Keywords: Drought, GIS, RS, NDVI

1. Introduction

There is increasing pressure on water resources as a result of global population growth, economic development and the adverse effects of land use and climate change. Any deficit or limitation in water supply will be most critical in drought periods, due to competing water needs. Drought is considered by many to be the most complex but least understood of all natural hazards, affecting more people than any other hazard [1]. Drought affects virtually all climatic regions and more than one half of the earth is susceptible to drought each year [2, 3]. Therefore, in order to design and manage water resource schemes, it is essential to monitor the spatial and temporal variability of droughts.

Traditional methods of drought monitoring rely on rainfall data, which are limited to the network of stations and incomplete climate data, often inaccurate caused a human error or crash equipment and most importantly, difficult to obtain data in near real time either spatially or temporally. In this way, remote sensing technology has greatly enhanced our ability to monitor and manage the natural resources, especially in the areas of water resources. The satellite or remote sensing techniques can be used to monitor the current situation- before, during or after disaster. They can be used to provide baseline data against which future changes can be compared while the GIS techniques provide a suitable framework for integrating and analyzing the many types of
A drought index assimilates thousands of data on rainfall, snow pack, stream flow and other water-supply indicators into a comprehensible picture. There are several indices that measure how much precipitation for a given period of time has deviated from historically established norms. Among various drought indices, the Palmer Drought Severity Index (PDSI) [4], Crop Moisture Index (CMI) [5], Standardized Precipitation Index (SPI) [6], and Surface Water Supply Index (SWSI) [7] are used extensively for water resources management.

NDVI has been extensively used for vegetation monitoring, crop yield assessment, and drought detection [8, 9]. NDVI is defined as (NIR-RED) / (NIR+RED). Where, NIR and RED are the reflectance in the near infrared and red bands. NDVI is a good indicator of green biomass, leaf area index, and patterns of production[9,10]. The output of NDVI is a new image file/layer. Values of NDVI range from -1.0 to +1.0. Higher values of NDVI signify a larger difference between the red and near infrared radiation recorded by the sensor - a condition associated with high photo synthetically active vegetation. Low NDVI values mean there is little difference between the red and NIR signals. The relationship between NDVI and rainfall is known to vary spatially, notably due to the effects of variation in properties such as vegetation type and soil [11].

2. Methodology

Study area

The study area taken is a part of the Jamnagar district of Gujarat between latitude 22°19′46″N to 22°46′01″N and longitude 70°20′56″E to 70°47′34″E. The area of study area is found to be 2341 km². The region of Jamnagar experiences a tropical climate, much similar to the rest of Gujarat.

Flow chart of image processing and NDVI calculation

Acquisition (download) of satellite images from archives of Landsat [Source – GLCF]
↓
Processing of Landsat images
↓
Verifying Geo-referencing of image by Google earth
↓
Layers stacked in ERDAS Imagine
↓
Cropping of interested part of the image considering AOI
↓
Classification of vegetation index i.e, NDVI
↓
Classifying different land cover representing different drought pattern

Figure 1. Map Showing Study area.

Figure 2. Satellite Image of Study Area.

Figure 3. Satellite Image of Study Area.
Figure 4. Satellite Image of Study Area.

Figure 5. NDVI Image for 17/10/1977 Imaginary.

Figure 6. NDVI Image for 02/11/1990 Imaginary.

Figure 7. NDVI Image for 18/10/1999 Imaginary.

Table 1. Analysis of NDVI Obtained From 1990 Imaginary.

| Range of digital number value | Vegetation class | Area under each group (km²) | Percentage of total area |
|------------------------------|------------------|-----------------------------|--------------------------|
| 0 - 100                      | 0 - 5 %          | 70.09                       | 2.99                     |
| 101 - 150                    | 5 - 20 %         | 1770.08                     | 75.61                    |
| 151 - 200                    | 20 - 35 %        | 438.81                      | 18.74                    |
| 201 - 255                    | > 35 %           | 62.07                       | 2.65                     |
|                              |                  | 2341.05                     | 100.00                   |

Table 2. Analysis of NDVI Obtained From Imaginary.

| Range of digital number value | Vegetation class | Area under each group (km²) | Percentage of total area |
|------------------------------|------------------|-----------------------------|--------------------------|
| 0 - 100                      | 0 - 5 %          | 85.10                       | 3.64                     |
| 101 - 150                    | 5 - 20 %         | 715.90                      | 30.58                    |
| 151 - 200                    | 20 - 35 %        | 1474.30                     | 62.98                    |
| 201 - 255                    | > 35 %           | 65.74                       | 2.81                     |
|                              |                  | 2341.05                     | 100.00                   |

Table 3. Analysis of NDVI Obtained From 1990 Imaginary.

| Range of digital number value | Vegetation class | Area under each group (km²) | Percentage of total area |
|------------------------------|------------------|-----------------------------|--------------------------|
| 0 - 100                      | 0 - 5 %          | 85.10                       | 3.64                     |
| 101 - 150                    | 5 - 20 %         | 715.90                      | 30.58                    |
| 151 - 200                    | 20 - 35 %        | 1474.30                     | 62.98                    |
| 201 - 255                    | > 35 %           | 65.74                       | 2.81                     |
|                              |                  | 2341.05                     | 100.00                   |

Figure 8. Area under different vegetation class in different year.
3. Discussions

Satellite images of 1977, 1990 and 1999 were downloaded and cropped according to area of interest as shown in Figure 2, 3 and 4 respectively. The satellite image of 1977 is of database MSS having resolution of 57 m, image of 1990 is of database TM (Thematic Mapper) having resolution of 28.5 m and 1999 having database ETM + (Enhanced Thematic Mapper +) having resolution of 15 m. All the three images were taken during post monsoon period.

NDVI images generated from satellite images of year 1977, 1990 and 1999 respectively using ERDAS-Imagine as shown in Figure 5, 6 and 7 respectively. On the basis of range of digital number, NDVI was classified into four groups. Table 1, 2 and 3 showing area covered by different vegetation class resulting from images of year 1977, 1990 and 1999 respectively. Figure 8 showing the comparative analysis of NDVI images results obtained. It can be observed that in year 1990, there was more area having higher percentage of vegetation. From comparison of annual rainfall of different years it was found that it was more in 1990 among three. It was therefore concluded that temporal variations of NDVI are closely linked with precipitation. The NDVI increases with increase in precipitation.

4. Conclusions

The ground meteorological data and remote sensing as well GIS technologies significantly contributes in the vegetation drought analysis.

The main objective of the paper was to identify the relationship between rainfall and NDVI and to see how appropriately drought risk areas can be delineated by integration of satellite, meteorological and other ancillary data. It was concluded from the study that the temporal variations of NDVI are closely linked with precipitation and there is strong linear relationship between NDVI and precipitation.

Unavailability of data at regular interval of time in public domain was found one major problem during the study. The resolution of data available is also coarse. The analysis will be better if data is available at regular interval of time and is appropriate drought risk areas can be delineated by integration of satellite, meteorological and other ancillary data.

During present study it was tried to correlate the NDVI with precipitation. If this analysis is performed for a large number of images for continuous years, it is convenient to find the recurrence of drought. On the basis of which the delineation of major areas facing drought risk and thereby management plans can be formulated by the government authorities to cope with the disastrous effects of this hazard. The further research can be use of the Vegetation Condition Index (VCI) to further separate regional NDVI variation from geographical contributions in order to assess regional drought impacts. Finally, all drought-prone nations should develop national drought policies and preparedness plans that place emphasis on risk management rather than following the traditional approach of crisis management, where the emphasis is on reactive, emergency response measures.

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