

Meteorological Drought in Three Woredas of Eastern Harerghe and Shinile Zones, Ethiopia

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Abstract

Climate has always been dynamic affecting natural systems through the consequence of its variability and change. Agriculture is the most vulnerable and sensitive sector that is seriously affected by the impact of manifested through rainfall variability and recurrent drought. In dry land semiarid areas of Ethiopia, including in the study areas, farmers inhabiting the area experience extreme temporal and spatial variability of rainfall in cropping season with frequent and longer dry spells. This makes them vulnerable to the risk of Meteorological Drought (MD). Thus, in order to adapt and/or mitigate the impact of MD, MD assessment has to form one dimension of research to be done whereas the use of MD indices and different soft wares provides wide scope in drought risk detection and mapping. Consequently, this study was conducted in Eastern Harerghe and Shinile Zones with the objective of assessing MD risk and preparing MD using Rainfall (RF). To assess and examine spatiotemporal variation of seasonal MD patterns and severity, a drought index namely, Standard Precipitation Index (SPI) anomalies are applied. To characterize temporal and spatial MD, rainfall data for a period of 36 years from (1980-2016) were utilized. The time of the seasonal transitions can also be an important control on drought, the result derived from index the annual trends of RF increased in Fedis; whereas, in Deder and Shinile woredas, it decreased with the time. The spatial distribution of drought frequency and its climate trend as well as different drought grades were obtained by employing the inverse distance weighting (IDW) interpolation method in ArcGIS 10.3.1 platform. In this study, the MD indices used to quantify the drought severity, frequency, trends, spatial pattern and its intensity. So, the annual high drought frequency mainly concentrated in the Eastern some areas of Fedis. From South to West part of Deder were observed slight to moderate drought. From slight to severe drought was largely experienced in shinile more than others study areas. Generally, the result revealed that SPI anomaly express decreased with the time over the study areas. Thus, the result indicated that the areas are classified into slight, moderate and severe MD risk zone. Thus, the MD study can be useful to guide decision making process in drought monitoring and to reduce the risk of drought on agricultural production and productivity.

Keywords: MD (Meteorological Drought) • Meteorological drought indices • Rainfall and geographical information system

Introduction

Background

Climate has always been dynamic, and affects natural systems through the consequences of climate variability and climate change [1]. Because of climate change and variability, drought has become a recurrent phenomenon in several countries across the globe. It is manifested in erratic and uncertain rainfall distribution in rainfall dependent farming areas, especially in arid and semi-arid ecosystems. Frequent and severe drought has become one of the

most important natural disasters in sub-Saharan Africa and often results in serious economic, social, and environmental crises [2]. It is marked by the creation of uncertain agricultural economics [3]. Droughts are a type of environmental disaster that occur in every climatic zone, no matter if high or low rainfall area. Droughts are usually divided into four different stages. The first stage is a meteorological drought which is defined as the cumulative lack of precipitation in a region over a prolonged period of time. Second a lack of water in the soil causes an agricultural drought, which usually is accompanied by reduced crop yields or shortage of soil moisture. Third a hydrological drought is characterized by a shortage of water in surface or underground water resources, it can be visible through dried up rivers and lakes [4]. Finally, a socio-economic drought is apparent when the supply of good water does not meet the demand. This model is currently operational as yield monitoring and forecasting tools for region wide food security analyses in drought prone countries in Sub-Saharan Africa [5]. In addition, Standard Precipitation Index (SPI) is meteorological drought index used to quantify the impact of rainfall deficit on soil moisture on which it responds to precipitation anomalies on a relatively short scale. In general, applications of satellite based vegetation, crop performance and climate derived drought indices become the feasible means and acceptable decision tools for timely action against the negative consequences of meteorological drought. Thus, Assessment of meteorological drought by remote sensing or ground data is believed to be very helpful for providing countermeasures as early as possible.

GIS and Remote sensing techniques are being widely used for analysis of drought monitoring and assessment using various drought indices [6-9]. Ethiopia is a sub-Saharan country that has been affected by frequent drought. Millions of lives have been lost because of recurring droughts in the past several decades. Due to climatic changes, drought occurs every two years in different parts of Ethiopia [10, 3]. In addition, the drought recurrence cycle shortens over time while the affected area is widening, impacting additional parts of the country that were once unaffected. In order to respond to the effects of drought, Ethiopia has been conducting drought assessment and monitoring missions. The Ethiopian National Meteorological Services Agency (NMSA) classified three seasons per year in Ethiopia: rainy season (June to September), dry season (October to January) and short rainy season (February to May) [10]. Since agricultural activities in the dry land semi-arid areas in general and study area, Fedis, Deder and Shinile woredas in particular, are influenced and controlled by seasonal rain, the study of meteorological drought analysis is carried out season-wise using climate based drought index (SPI). These index as proxies of drought, and crop and/or vegetation health condition are vital in drought assessment and hence, they are applied in this study. Problem Statement is Ethiopia's climate is influenced by general atmospheric and oceanic factors that affect the weather system and climatic factors such as drought has significant impact on agricultural output [11]. This makes Ethiopia vulnerable to climate variability. Climate variability over the last three decades resulted in drought and famines in Ethiopia and several other countries in Africa [12]. Due to greater reliance on climate sensitive sectors such as agriculture, Ethiopia is vulnerable to the risk of climate variability and change such as MD. MD is a period of abnormally dry weather sufficiently for the lack of precipitation to cause a serious hydrological imbalance and carries connotations of a moisture deficiency [13]. In Eastern Ethiopia, the major impact of MD is manifested through rainfall variability and recurrent drought due to climate change.

The amount and distribution of rainfall in these crops growing season is also very erratic, indeed making crops vulnerable to moisture deficit. Agriculture is one of the most vulnerable sectors to MD impact. The impact is redundantly stronger in the study areas in particular, where agriculture is truly important for the daily subsistence, and where adaptive capacity is low. Agriculture is the most vulnerable and sensitive sector that is seriously affected by the impact of MD. MD is currently more pronounced in the study area. Farmers inhabiting in the study area are exposed to the impact of MD that resulted from the current climate change. This is due to the fact that most of farmers' cultivars (long duration crop variety) can no longer resist the existing MD and intra seasonal dry spell [10]. As a result, most of the cultivars have gone obsolete, and yield reduction and crop failure, death of livestock becomes common [14]. Therefore, the aim of this research was to study the impact

of Meteorological-Drought due to precipitation performance, to assess quantitatively climate anomalies in terms of severity, intensity (magnitude of drought), duration, frequency and spatial extent) in the introduction explained areas. Generally, studying of the impact of Meteorological drought (drought severity, intensity, frequency and duration) on crops can be useful in this regard for the decision making process on drought monitoring and to avert its consequences on agricultural production and productivity for study areas. The main objective of the study was to investigate the spatial-temporal occurrence of meteorological drought in the study areas. Moreover, this study generated information about impact of MD on crops and facilitates exchange of this information among societies, field experts, policy makers and researchers. The study tried to investigate Meteorological drought spatial-temporal characteristics and its impact. The study therefore, would to provide significant recommendation for study areas.

Data and Methodology

Topography Eastern Harerghe and Shinile Zones are characterized by diverse topographic features in which high mountainous and deeply incised canyons and gorges, valleys and plateaus with steep slopes dominate its most parts [15]. Because of the diverse topography, the study area experiences different climatic conditions that range from hot and arid lowlands in the eastern part to cold and humid highlands in the western part Mentioned by the following Table 1.

Table 1. Classification of agro-ecological zones based on altitude.

Elevation (meter above sea level)	Traditional zone	Description
>3000	Wurch	Cold Highlands
2500-3000	Dega	Cold, Humid highlands
1500-2500	Weina Dega	Temperature, Cool Sub humid highlands
300-1500	Kolla	Warm, Semi-Arid Lowlands
<300	Bereha	Hot and Hyper-Arid Lowlands

Climate

The climate of study areas in Eastern Harerghe Zone: Fedis, Deder woreda and Shinile Zone shinile woreda studied by semi-arid and sub-humid climate based on the moisture index classification of climate. The long term average seasonal rainfall that the area receives ranges from 458 to 518 mm and according to NMA study in 1996, the study areas are characterized by bimodal rain fall distribution (those were spring and summer). The warmest temperatures are recorded during February to May, while the coldest was recorded during October to January. High temperatures between 30-35oc are recorded in low altitude areas, however in highland areas the average temperatures are between 20-22oc and NMA website (map-room, www.ethiomet.gov.et) [16].

Data collection and analysis

To achieve the objective of this study the following procedures were followed: The study used ArcGIS version10.3.1 software to characterize spatial Meteorological Drought by imputing (analysed) rainfall data processed by MS Excel, woredas' coordination (latitude and longitude) and woredas' base map

data. The study used MS excel 2010, Instant version 3.37 and XL Instat softwares to characterized temporal Meteorological Drought (i.e. to show rainfall trend, extent, frequency and duration in temporal characteristics of MD as output) by analysed rainfall and crop yield data; For this study, the study used a time series AVHRR satellite gridded and three stations rain gauges' rainfall data used to show meteorological–drought characters by helping indices such as SPI4 based (because they are used to assess meteorological season classification based on Lema Gofa studie. The data obtained from National Meteorological Agency Jijiga Branch Office (NMA JBO), AVHRR gridded rainfall data have 4km by 4km spatial resolution (satellite data) in monthly and seasonally time step basis data for the period of the main rainy season (June–September) and (Mar–may). This gridded obtained from NMA and Famine Early Warning System (FEWS-NET) archive website: (<http://iridl.ldeo.columbia.edu> GCM) about 37 years (from 1980-2016 G.C) that used to fill the missing year data. The objects in this case are defined based on attributes of their physical feature and drought is a geographic object. So to So to characterize meteorological drought, (Figure 1) the study used long year's meteorological data and different soft wares such as INSTAT 3.37 and ArcGIS 10.3.1 techniques, MS Excel 2010, and these woreda base map. Finally, this research is to assess impact of Meteorological drought using Meteorological

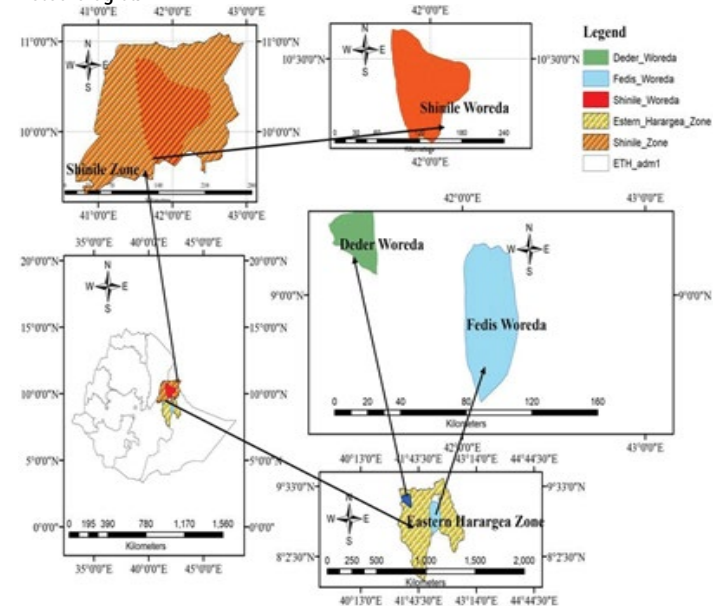


Figure 1. Map of the study areas (by using GIS version 10.3.1 Software).

Stations, satellite data and based on Meteorological drought indices and by helping different software packages (to assess quantitatively climate anomalies in terms of severity, intensity (magnitude of drought), duration, prediction and probability, frequency, temporal and spatial extent of drought).

Data sources and collection method

For this study, the daily historical meteorological data collected from Ethiopian National Meteorological Agency (NMA) for a period of 36 years (1980-2016). These include daily rainfall; gridded (Station+satellite) RF on the study area. Three meteorological weather stations shown above in Table 2 selected for the study on the basis of spatial representative of the zone.

Table 2. Representative Meteorological stations for study areas.

Study Areas	Latitude (N)	Longitude (E)	Atitude (m)	Area Coverage (in Km ²)	Annual RF (in mm)	Annual Mean Tem (in C°)	
Shinile	9° 41' 37.2"	41° 50' 92.8"	41° 50' 92.8"	1029	0.5	639	25.7
Deder	9° 19' 79.5"	41° 26' 71.3"	41° 26' 71.3"	2144	0.03	1008.1	16.4
Fedis	9° 07' 94.4"	42° 04' 49.7"	42° 04' 49.7"	1699	0.11	526.1	19.5

Data analyses

To characterize temporal and spatial Meteorological-Drought, its impact with rainfall; monthly rainfall data for a period of 36 years from (1980-2016) rearranged in a monthly, seasonally and annually time step in Microsoft excel 2010 spread sheet. For the study areas (Fedis, Deder and shinile), the trend and severity of monthly, seasonally and annually rainfall (i.e. maximum, minimum and mean rainfall). The rearranged data averaged over more than thirty years' time span and analyzed in order to examine the seasonal, monthly, seasonal and annual Meteorological-drought characteristics by using the following indices and different soft wares. The parameters used in this study are used described in the following sections.

Standard Precipitation Index (SPI)

SPI is an index that is developed to quantify precipitation deficit at different time scales, and can also help assess drought severity of MD by helping MS Excel and by using ArcGIS version 10.3.1 software analysed spatial pattern of drought. The SPI is simplicity (only rainfall data is required for calculation). SPI is also used to analyse the impact of rainfall deficiency on crop development. Different time scales reflect the impact of drought on the availability of the different agricultural crops. According to Edwards and McKee in 1997, the mean SPI for the location and desired period is represented zero and also it represented water deficit or excess (i.e. positive SPI represents that water has been available to plants whereas negative SPI or rainfall deficiency is reflected on crop production through yield reduction). Yield reduction due to moisture deficiency is clearly explained in this issue. The formula of SPI is:

$$SPI = (X_{ij} - X_{im}) / \sigma \quad \text{.....(Equation 1)}$$

Where, X_{ij} is the seasonal precipitation and, X_{im} is its long-term seasonal mean and is its standard deviation. The final result of SPI based drought severity class are: >0=No drought, 0.0 to -0.99=Slight drought, -1.0 to -1.49=Moderate drought, -1.5 to -1.99= Severe drought, -2 and less, Very severe drought that mentioned by Gizachew in 2010 as followed Table 3.

Table 3. SPI based drought severity class.

SPI Value	Drought severity class
Above 0	No drought
0.0 to -0.99	Slight drought
-1.0 to -1.49	Moderate drought
-1.5 to -1.99	Severe Drought
-2 and less	Very Severe drought

So, it is used to see the duration, severity, frequency of meteorological-drought.

Characterizing the meteorological-drought

From the daily of long term rainfall data, the onset and cessation dates, trend analysis, and dry spell risk analysed. All these Meteorological-Drought indices identified different soft wares, such as INSTANT tools version 3.37, and SPI4 and the data captured by MS Excel 2010 soft wares. In order to examine the MD of study areas for more than a thirty years (1980-2016) climatological period Statistical tools like mean, standard deviation, coefficient of determination (R^2), Coefficient of Variability (CV) and p-value comparison used in INSTANT (v 3.37) statistical software. Coefficient of Variability used to classify the degree of MD (rainfall) events as less, moderate and high. When CV <20% it is less variable, CV from 20% to 30% is moderately variable, and CV >30% is highly variable. Areas with CV > 30% are said to be vulnerable to drought [17].

$$CV = \left(\frac{SD}{X} \right) \times 100, \text{ where } X = \frac{\sum X_i}{N} \text{ and}$$

$$SD = \frac{\sqrt{(X_i - X) \times 2}}{N - 1} \quad \text{.....(Equation 2)}$$

Where from equation (5) CV is the coefficient of variation, SD is standard deviation, is long year mean and N is total number of year during

observations, X_i is rainfall of each month or season. The coefficient of determination (R^2) also indicates the variation of the climate variable with time in the linear regression model which is defined as:

$$R^2 = SSR / SST$$

Or

$$1 - SSE / SST \quad \text{.....(Equation 3)}$$

Where, SSE is the Sum of Squared Error, SSR is the Sum of Squared Regression; SST is the Sum of Squared Total. R^2 measured how successful the fit was in explaining the variation of the data. P-value is a statistical measure that used to determine whether the hypotheses were correct. Usually, if the p-value of the dataset was below the pre- determined amount (say 0.05 which is the 95% confidence interval), the variability of the dataset has meaningful effect on the result [18].

Trend analysis

For trend analysis, the monthly and annually data series of the study areas weather station more than 30 years long year data analysed and test for the existence of significant trend by correlation and regression analysis using INSTANT software. To determine the trend of rainfall in the study areas, sequential plot time series analysis used. This can be mathematically denoted using the functional relationship below:

$$Y = f(i) \quad \text{.....(Equation 4)}$$

Where, Y is the value of the variable under consideration at time i.

Suppose that Y_i is the estimated value of the dependent variable Y_i at time i, then Y_i can be predicted from the simple linear equation.

The model can be written as

$$Y_i = \beta_0 + \beta_1 X_i + \varepsilon_i \quad \text{.....(Equation 5)}$$

Y_i is the estimated value of the dependent variable (maize yield); β_0 = the Y intercept measuring the value Y_i at time i, $\beta_1 X_i$ is the slope of the line measuring the change in the variable Y_i that results from a units change in time i, i=the independent variable (time) i=1, 2, 3, 4...30 and ε_i =Random error. It hypothesize that there is no trend in the amount of rainfall over time. Thus the null hypothesis will stated as; $H_0: \beta_1=0$. Variability of annual and seasonal rainfall will be assessed using CV and Analysis of Variance (ANOVA) techniques. Standard rainfall will plotted against time (in years) to visualize the time series variation of annual and seasonal rainfall.

The regression model for the study rain fall effect will be computed as:

$$Y = a + bx \quad \text{.....(Equation 6)}$$

Where

Y=the value of the dependent variable (maize and sorghum yield/ ha)

a=Y intercept

b=regression coefficients (each b represents the amount of change in Y (yield) for one unit of change in the corresponding x-value when the other x values are held constant; x=the rainfall independent variables (i.e. rainfall) and e=the error of estimate or residuals of the regression.

Pearson Correlation coefficient (r) analyses will be used to analyse the correlation between times (expressed in year) and rainfall characteristics, value of r ranges between -1 to +1, a correlation coefficient close to +1 indicates a strong positive correlation, a correlation coefficient close to -1 indicates a strong negative correlation similarly a correlation coefficient of 0 indicates no correlation.

Results and Discussion

Spatial and temporal patterns of drought conditions using indices

Standard Precipitation Index (SPI): Drought characteristics included various drought conditions, such as duration, severity, and intensity. The SPI was determined based on seasonal classification of Ethiopian which mainly include Belg (Feb-May), Bega (Oct-Jan) and Kiremt (Jun-Sep) [10,19]. The results of SPI4 analysis are given below in Table 4 and Figure 2. The point RF

was used to talk worda based because the NMA station RF data represent about fifty Kilo meter radius coverage.

Table 4. Anova of study areas.

St. Name	MK Slope	Season	Kendal's Tau	Sen's slope	P Value
Shinile	-111	Belg	-0.19	-0.075	Sig
	-76		-0.12	-0.841	Sig
	-144		-0.18	-1.919	Sig
	-101		-0.16	-0.767	Sig
	161	Kiremt	0.26	0.538	Non- Sig
	188		0.3	1.58	Non- Sig
	73		0.12	0.737	Sig
	23		0.04	0.206	Sig
	-197	Belg	-0.34468	-0.82857	Non- Sig
	-63		-0.10734	-0.74	Sig
Deder	-106		-0.17921	-1.5167	Sig
	-117		-0.20144	-1.55333	Sig
	90	Kiremt	0.154003	0.447815	Sig
	6		0.010179	0	Sig
	-52		-0.08791	-0.75	Sig
	-53		-0.0903	-0.26496	Sig
	4	Belg	0.007057	0	Sig
	49		0.0795	0.1579	Sig
	-65		-0.10443	-0.6057	Sig
	190		0.308521	0.126798	Non- Sig
Fedis	123	Kiremt	0.197605	0.815302	Sig
	170		0.271135	1.432248	Non- Sig
	165		0.263372	1.837037	Non- Sig
	5		0.008007	0	Sig

Sig: Significant, Non-Sig: Non significant, MK: Mann Kendall

The analysis of SPI4 for Shinile worda showed that drought has occurred at different level from 1980 to 2016 rainy season. In the below Figure 2, the drought happened in the years 1980, 1981, 1984, 1985, 1988, 1989, 1990, 1991, 1995, 1999, 2000, 2002, 2003, 2004, 2005, 2008, 2009, 2011 and 2015. There were very severe droughts years and with their more intensity and duration [20]. These years were 1984, 2000, 2005, 2009, 2011 and 2015 those were explained with their SPI values that range from -2 to -1. The results indicated that during those years, there was rainfall deficit in the growing season and therefore, there was the worst dry season. SPI can be used to identify not only drought years but also wet years. In this regard, using SPI analysis, wet years were identified. As the maps were shown in the Figure 2, the years 1982, 1983, 1986, 1987, 1992, 1993, 1994, 1996, 1997, 1998, 2001, 2006, 2007, 2010, 2012 and 2013 were very wet years in Shinile worda, and based on SPI4, among these years the extreme wettest years were 1983, 1997, 1998, 1996 and 2010. Besides analysis of the 36 years (1980-2016) annual RF showed a coefficient of variation ranging from 70.4%-157.2%.

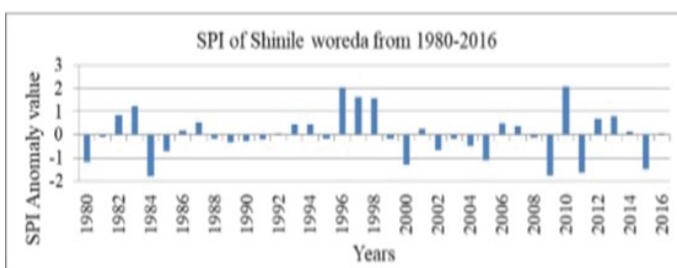


Figure 2. SPI RF anomaly of Shinile worda for the period of 1980-2016.

The analysis of SPI4 based for Deder worda, from Figure 3, the years 1984, 1985, 198, 1987, 1991, 1994, 2004, 2005 and 2007 to 2015 were droughts. There were very severe droughts with their SPI values ranges from -1.0 to -1.5 those were the years 1984, 1994, and 2007 to 2013. The results indicated that during those years there was rainfall deficit in the growing season and therefore, there were the worst dry years [21]. Considering of the most severe droughts years above mentioned period was a better illustration of the drought extent and intensity in this area. So based on this index, the extreme wet years were 1999, 1998, and 1997 with SPI values greater than 1.0. Besides analysis of the 36 years (1981-2016) annual RF showed a coefficient of variation ranging from 70.8%-139.0%. This was highly variation of RF.

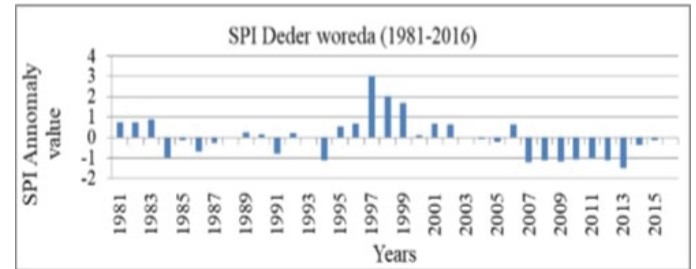


Figure 3. SPI RF anomaly of Deder worda for the period of 1981-2015.

The result of the analysis of SPI4 for Fedis worda is given in Figure 4. As it can be seen, the droughts occurred in the years 1980, 1982-1986, 1990, 1991-1993, 2005, 2007, 2010, 2011, 2012 and 2015. There were very severe droughts in the years 1980, 1982-1985, 1986, 2010 and their SPI values range from -1.5 to -2.0. The results indicated that those years were rainfall deficit in the growing season and therefore, there were the worst dry seasons in Fedis worda. Considering of the most severe droughts above mentioned period were the better illustration of the droughts extent and intensity in Fedis worda. The years 1988, 1989, 1995, 1996, 1997, 2009 and 2014 were also extreme wet with SPI values greater than 1.0. Besides analysis of the 36 years (1981-2016) annual RF showed a coefficient of variation ranging from 75.6%-202.7%.

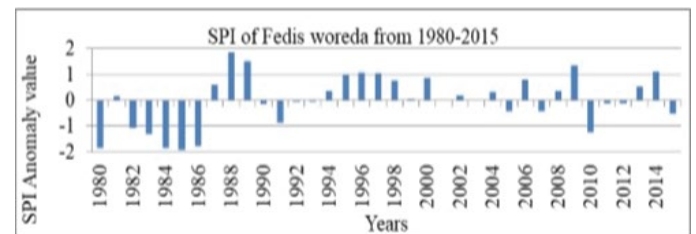


Figure 4. SPI RF anomaly of Fedis worda for the period of 1980-2016.

During Kiremt season (Jun-Sep), in the Figure 5, based on SPI4 for Deder in the years 1984, 1987, 1991, 2010 and 2013 were found to be the extreme drought; whereas, during this season the years 1983, 1988, 1998, 1999, 2000, 2001 and 2002 were found to be extreme wet and based on vegetation (NDVI index) trend data these years were good production years. As the Figure 5 indicates, most years have shown more frequent drought than wet years and also drought occurrence was frequent in 1980s and 2010s; this indicates the earliest time is returned now and the RF performance is shown decreasing trend.



Figure 5. Kiremt Season (Jun-Sep) RF SPI of Deder worda for the period of 1981-2015.

During Kiremt season (Jun-Sep) in the Figure 6, based on SPI4 for Fedis woreda in the years 1980-1986, 1990-1993, 2001,2005, 2007, 2010, 2012 and 2015 were shown droughts (the SPI4 values range from -0.1 to -2.0). Among these years, 1980, 1984, 1985 and 1986 were very severe droughts (SPI4 values range from -1.6 to -2.0); whereas, the years 1987-1989, 1994-2000, 2002, 2006, 2008, 2009, 2011, 2013 and 2014 were wet (the SPI4 values range 0.3 to 1.6). Among these years, 1988 and 2009 were good cropping years because these years were found to be extreme wet years than others [22]. As the below Figure 6 has shown frequency of droughts years were more than wet years and also drought occurrence was frequent in the first years (1980's).



Figure 6. Seasonal (Jun-Sep) RF SPI of Fedis woreda for the period of 1980-2015.

During Kiremt season (Jun-Sep) in the Figure 7, based on SPI4 for Shinile woreda, in the years 1980-1982, 1984-1987,1989-1991, 1993, 1995, 1997, 2002, 2004, 2009 and 2015 were the droughts (the SPI4 values range from -0.2 to -2.1). Among these years, 1991 and 1987 were very severe droughts (the SPI4 values range from -1.6 to -2.1); whereas, the years 1984, 1996, 1998, 2001 and 2010 were extreme wet and these years were good cropping years because these years were more wet years than others. So, as the below Figure 7, based on SPI4 has shown frequency of drought years were less than wet years and also drought occurrence was frequent in the first years (1980's).

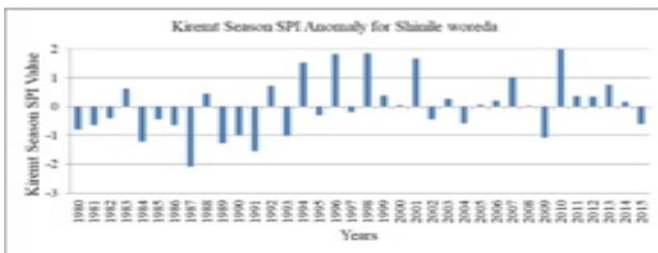


Figure 7. Kiremt Season (Jun-Sep) RF SPI of Shinile woreda for the period of 1980-2015.

As it can be seen from Figure 8, the SPI4 values for Belg season (Feb -May), ranged from 0.1 to -1.6 for Shinile woreda, during the years 1980, 1984, 1985, 1988, 1992, 1994, 1999-2005, 2007, 2009, 2011-2013 and 2015. These values indicate drought of different levels. Based on the classification given by equation 3.3, SPI values between -1.5 and -1.6 are considered as severe droughts. In line with this, the years 2000, 2009, and 2011 are classified severe droughts [23]. The earliest years (2000's) were characterized with more frequent drought events whereas the years 1981, 1983, 1986, 1987, 1996 and 2010 were wet and there were good cropping season.



Figure 8. Belg season (feb-may) RF SPI of shinile woreda for the period of 1980-2015.

During Belg season (Feb-May), in the Figure 9, based on SPI4 for Deder woreda, in the years 1984, 1986, 1988, 1992-1994, 2000-2005, 2007-2009, 2011-2015 were shown drought. Among these years, the years 1988, 1994, 2007, 2009, 2011-2013 were shown from moderate to severe drought (the SPI4 values range from -1.0 to -1.6). The recent years were more frequent the

drought years whereas from the years 1981-1983, 1985, 1987, 1989-1991, 1995- 1998, 2006 and 2010 were wet (the SPI4 values range from 0.1 and 3.3) [24]. Among these years, in 1997 was the wettest year and expected good cropping season.



Figure 9. Belg Season (Feb-May) RF SPI of Deder woreda for the period of 1981-2015.

During Belg season (Feb-May), in the Figure 10, based on SPI4 for Fedis woreda, in the years 1980, 1982-1986, 1991,1999, 2001, 2002, 2005, 2008, 2010, 2011 and 2015 were shown droughts (the SPI4 values range from -0.1 to -1.5). Among these years, the years 1980, 1982,1986, 1999, 2008, 2010 and 2015 were shown moderate droughts (the SPI4 values range from -1.0 to -1.5) specially 1984 were severe drought [25]. The earliest years (1980's) were more frequent the drought years; whereas, in the Figure 10, the years 1981, 1988-1990, 1995-1998, 2000, 2003, 2004, 2007, 2009 and 2014 were wet (the SPI4 values range from 0.1 and 2.0). Within line this, 1998 and 1989 were found to be the extreme wet years and there were expected good cropping season.



Figure 10. Belg Season (Feb-May) RF SPI of Fedis woreda for the period of 1980-2015.

So, the implication of the above result indicated that the drought pattern such as drought frequency, severity and intensity explained by using SPI4 index in study areas. It used to know the drought years, its frequency time and its magnitude in study areas.

RF trend analysis in the study areas

As can be seen in Figure 11, annual total rainfall trend in Fedis woreda for the period 1980 to 2016 was positively increased with the time. Based on the result in Figure 11, during the above mentioned period, the highest amount of RF was 923.3 mm in the year 1988; whereas, the lowest total amount of rainfall was 89.4 mm in the 1985 and also the mean total RF was 514.7 mm. The trend of total rainfall amount was increased with the time (see Table 5).

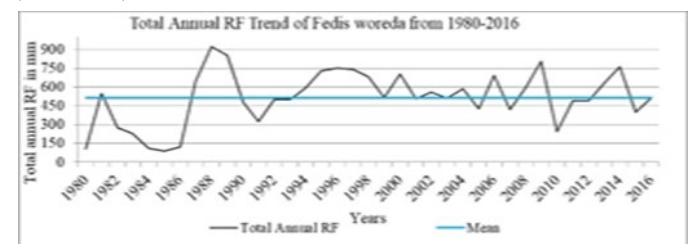


Figure 11. Total Annual RF Trend of Fedis woreda for the period of 1980-2016.

Simple linear regression between rainfall and years starts from 1980-2016 in Fedis woreda.

Table 5. ANOVA Fedis Woreda RF

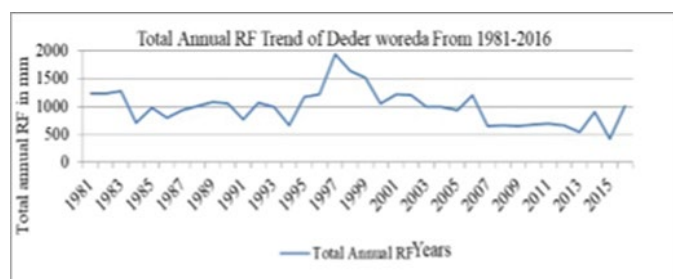
Significance							
	Df	SS	MS	F	F	-	-
Regression	1	433.7415	433.7415	4.011605	0.0529811	-	-
Residual	35	3784.289	108.1217	-	-	-	-
Total	36	4218	-	-	-	-	-
	Coeff	SE	t stat	P-value	Lower 95%	Upper 95%	Lower 95.0%
Intercept	1989.71	4.48	444.37	0	1980.62	1998.8	1980.62
RF Annually	0.02	0.01	2	0.05	0	0.03	0

Table 6. Anova deder woreda.

Significance							
	Df	SS	MS	F	F	-	-
Regression	1	357.2	357.2	3.4426	0.0722226	-	-
Residual	34	3527.8	103.759	-	-	-	-
Total	35	3885	-	-	-	-	-
	Coeff	SE	t stat	P-value	Lower 95%	Upper 95%	Lower 95.0%
Intercept	1989.71	4.48	444.37	0	1980.62	1998.8	1980.62
RF Annually	0.02	0.01	2	0.05	0	0.03	0

Fitted equation: $y = 0.0161 \cdot RF + 1989$ R-squared: 1.0

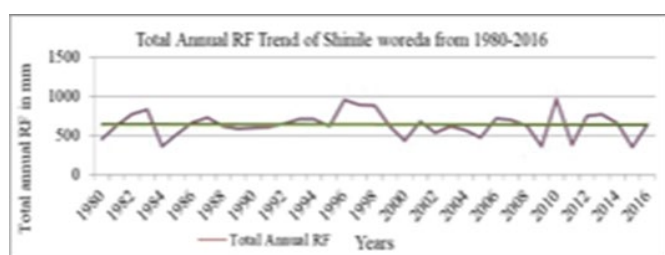
Whereas, the annual total rainfall trend in Figure 12, for Deder woreda for the period of 1981 to 2016 was decreased and the highest total RF recorded in the year 1997 with the value 1933.7 mm but the minimum total amount of rainfall recorded in 1985 with the value 425.0 mm. The mean total RF was 992.95 mm (see Table 6).

**Figure 12.** Total Annual RF Trend of Deder woreda for the period of 1981-2015.

Simple linear regression between rainfall and years starts from 1980-2016 in Deder woreda.

Fitted equation: $y = -0.0106 \cdot RF + 2009.2$ R-squared: 1.0.

As shown in Figure 13, the annual total rainfall trend in Shinile woreda for the period of 1980 to 2016 decreased and the maximum total rainfall was 965.9mm in 2010; whereas, the minimum total amount of rainfall was 348.6mm in 2015 and the mean total RF was 639.0mm. The study showed that the trend of total rainfall amount decreased with the time. Statistically, significant all trends were detected for precipitation on annual based (see Table 7).

**Figure 13.** Total Annual RF Trend of Shinile woreda for the period of 1981-2013.

Simple linear regression between rainfall and years starts from 1980-2016 in Shinile woreda.

Fitted equation: $y = -0.0028 \cdot RF + 1999.8$ R-squared: 1.0

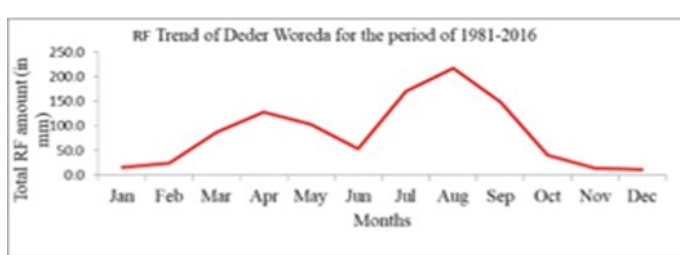
Additionally, all drought years with seasons by using SPI index were explained (see Table 8).

Monthly total RF trend analysis

As can be seen from Figure 14, the monthly total rainfall trend of Shinile woreda is decreased with time. The maximum RF recorded in the month of April and August during Belg season (Feb-May) and during Kiremt season (Jun-Sep) with value average 107.9 mm and 115.3 mm respectively. The minimum total RF recorded in December with value 12.4 mm and the monthly mean value was 53.3 mm.

**Figure 14.** Monthly Total RF Trend of Shinile woreda for the period of 1980-2016.

As can be seen from Figure 15, the monthly total rainfall trend in Deder woreda is slightly decreased with time increased and the highest total RF is also recorded on the month April and August. The highest value on average recorded 128.2 mm and 216.7 mm during Belg season (Feb-May) and Kiremt season (Jun-Sep) respectively. The minimum total RF is recorded 11.0 mm in December and the monthly mean value was 84.4 mm.

**Figure 15.** Monthly Total RF Trend of Deder woreda for the period of 1981-2016.

As can be seen from Figure 16, the monthly total rainfall trend, in Fedis woredas can be shown the increased trend with the time; and the maximum total RF recorded 128.2 mm during belg season (Feb-May) and 216.7 mm

Table 7. Anova shinile district.

Significance								
	Df	SS	MS	F	F			
Regression	1	6.908	6.908	0.0574	0.812028	-	-	-
Residual	35	4211.092	120.317	-	-	-	-	-
Total	36	4218	-	-	-	-	-	-
	Coeff	SE	t stat	P-value	Lower 95%	Upper 95%	Lower 95.0%	Upper 95.0%
Intercept	1999.781	7.65	261.403	0	1984.251	2015.312	1984.251	2015.312
RF Annually	-0.003	0.012	-0.24	0.812	-0.026	0.021	-0.026	0.021

Table 8. All drought years with seasons by using SPI index.

Drought years in study areas using different indices								
S.N	Period analyzed	Parameters	Study areas	Season	Drought Years	Index values	Index	
1	1980-2016		Fedis	Kiremt	1980-1986, 1990-1993, 2001, 2003, 2005, 2007, 2010, 2012& 2015			
				Belg	1980, 1982-1986, 1991, 1999, 2001, 2002, 2005, 2008, 2010 & 2015			
2	1980-2016	RF	Shinile	Kiremt	1980-1982, 1984-1987, 1989-1991, 1993, 1995, 1997, 2002, 2004, 2009 & 2015	< 0, 0 to -0.99, -1		
				Belg	1980, 1984, 1985, 1988, 1992, 1994, 1999-2005, 2007, 2009, 2011-2013 & 2015	to 1.49, -1.3 to -1.9, < -2.0	SPI	
3	1981-2016		Deder	Kiremt	1984-1987, 1989-1991, 1994, 2007-2015			
				Belg	1984, 1986, 1988, 1992-1994, 2000-2005, 2007-2009, 2011-2015			

during Kiremt season (Jun-Sep). The highest value on average was recorded in the month of April and August. The minimum total RF recorded 11.0 mm in December and the monthly mean value was 84.4 mm (see Table 5).



Figure 16. Monthly Total RF Trend of Fedis woreda for the period of 1981-2016.

Spatial pattern of drought severity, frequency and trend

Spatial pattern of drought severity, frequency and trend for fedis: Drought characteristics various characteristics can be used to describe droughts [25]. They Uses three characteristics, intensity, duration and spatial coverage. Intensity describes the magnitude of deficiency while duration is defined as the time of said deficiency, and spatial coverage the area experiencing a deficiency at a certain point of time. Spatial coverage and intensity changes with time as the drought develop. Severity can be described as a product of duration and intensity, and is frequently associated with the vulnerability of an area to drought [26]. Severity can also be defined as the socio-economic impact of droughts, often expressed in economical values of lost production or other costs affiliated with droughts. Finally, severity is also used in indices to distinguish the different phases. The spatial distribution of drought frequency and its climate trend as well as different drought grades were obtained by employing the inverse distance weighting (IDW) interpolation method in ArcGIS 10.3.1 platform [27]. There are very significant spatial differences of meteorological drought frequency from 1980 to 2016 in the study areas [28]. As it can be seen from Figure 17A, the annual spatial distribution of RF coverage increased both from western part of Fedis woreda that is large portion of woreda covered no drought from long term mean annual RF (470 mm), and from North to central part of Fedis was shown slight RF distribution and pocket areas RF was shown decreased or moderate drought the existence of some RF performance (Figure 17A). The annual spatial distribution of RF coverage was decreased or high drought frequency mainly concentrated in the Eastern some pocket areas of Fedis. From Figure 17B, the spatial pattern of most parts of the areas was slight decreased and the mean RF amount was (178 mm) from Northern to South parts of the District and also some pocket areas the RF coverage increased in the North Western; and the RF coverage was shown decreased in some pocket areas in the Eastern parts of the district [29]. From Figure 17C the spatial pattern of most parts of the areas the RF distribution was slightly decreased and the mean RF amount was (238 mm) in the North central parts of the District and also there was

increased in some pocket areas in the North Western and some pocket areas

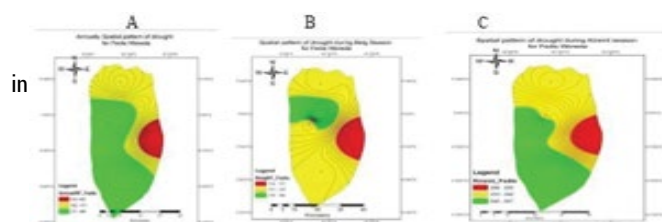


Figure 17. Spatial coverage of RF for Fedis A) Annual B) Belg C) Kiremt.

Spatial pattern of drought severity for fedis expressed in SPI index: As shown below from Figure 18A, the SPI index of annually (Jan-Dec) result most areas of the district was wet (no drought) in the South Western areas; from North to central part of areas was shown slight drought; whereas, some pocket areas in the Eastern part was shown the most drought (Moderate drought). From Figure 18B, most areas during Belg (Feb-May) was shown slight drought in the Northern and South Western parts of areas and some pocket areas in Eastern parts of Fedis was moderate drought; whereas, from central to Western parts of Fedis was shown wet (areas had no drought). From Figure 18C, most areas during Kiremt (Jun-Sep) also covered by slight drought in the Northern and central parts of Fedis and some pocket areas in Eastern parts of Fedis was moderate drought; whereas, in the South Western parts of Fedis was shown wet (areas had no drought).

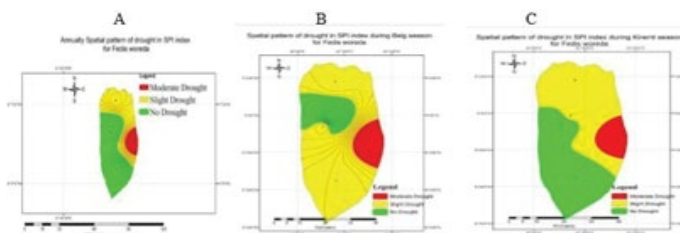


Figure 18. Spatial pattern of drought severity A) Annual B) Belg C) Kiremt for Fedis Woreda expressed in SPI index.

Spatial Pattern Drought Severity, Frequency and trend for Deder: As the result can be seen from Figure 19A, the annual spatial distribution of RF increased both from Eastern part of Deder woreda that is large portion of woreda (was shown wet) from long term mean of annual RF (754 mm); whereas, in the Western and central parts of Deder, the RF distribution slightly decreased; and

it also decreased in the Western tip and Southern tip. From Figure 19B, the Belg (Feb-May) the RF distribution was the same as annually RF distribution occurrence and the long term mean of Belg RF amount was 239 mm. From Figure 19C, the Kiremt (Jun-Sep), there was good RF performance in the South Eastern parts of Deder; whereas, from central to Western parts of Deder, there was slightly decreasing of RF performance and the long term mean of Belg RF amount was 456 mm.

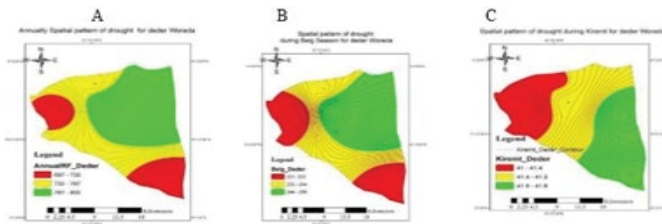


Figure 19. Spatial pattern drought severity A) Annual B) Belg C) of deder woreda.

Spatial pattern of drought for Deder expressed in SPI index: The study have been as the following figures by using SPI index anomaly most of the results was observed the same as the above observed Figures but for further explanation from Figure 20A, annually spatial pattern most of areas was shown no drought (low drought frequency) from Eastern to the central parts of areas; whereas, from South to West part of Deder were observed slight to moderate drought. From Figure 20B, Belg season spatial pattern in SPI index most of area was shown wet (no drought); from central to South and Western part of areas in SPI index were observed slight to moderate drought respectively. From Figures 20C, Kiremt season spatial pattern in SPI index most of areas was shown wet (no drought) Eastern to Central part of areas; from Western part of areas in SPI index was observed slight drought and moderate drought was observed in the Southern parts of areas.

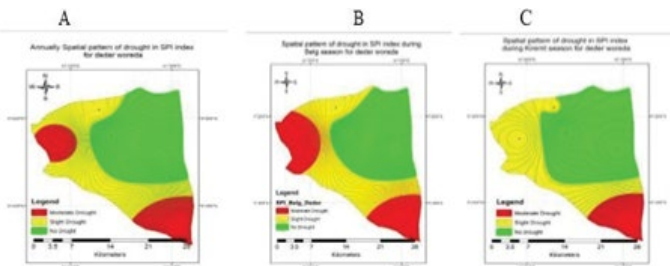


Figure 20. Spatial pattern of drought severity A) Annual B) Belg C) Kiremt for Deder Woreda expressed in SPI index

Spatial pattern drought severity, frequency and trend for Shinile: As the result can be seen from Figure 21A, the annual RF distribution expanded from North to central and Eastern part of Shinile woreda that is large portion of woreda covered shortage of RF (was shown dry) from long term mean of annual RF (639 mm); whereas; it was increased from central to South parts of Shinile (wet area). From Figure 21(A and C), the Belg (Feb-May) and the Kiremt (Jun-Sep) drought frequency were the same as annually RF distribution occurred and the long term mean of Belg and Kiremt RF amount were 258.9 mm and respectively.

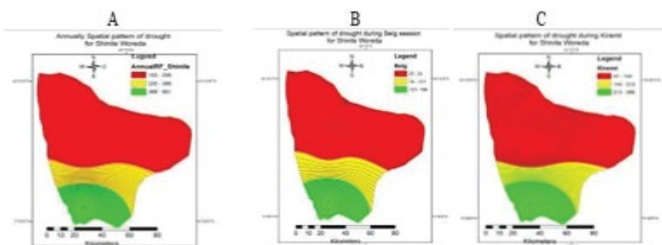


Figure 21. Spatial pattern of drought severity A) Annually B) Belg C) Kiremt of shinile woreda expressed in SPI index.

Spatial pattern of drought for Shinile woreda expressed in SPI index: The study have been as the following Figures, by using SPI index anomaly most of the results were observed the same as the above observed Figures but for further explanation from Figure 22A, annually spatial pattern small

pocket areas of Shinile woreda was shown no drought (low drought frequency) in the South parts of areas; whereas, from South to North tip part of Shinile were observed slight to moderate drought respectively. In this time large portion of areas was covered by moderate drought and more drought patterns during Kiremt and Belg seasons was shown similar to Figures 22(B and C).

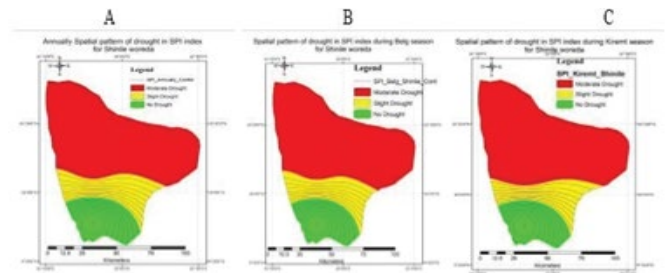


Figure 22. Spatial pattern of drought severity A) Annual B) Belg C) Kiremt for Shinile Woreda expressed in SPI index.

Summary

Climate has always been a dynamic, and it can affect natural systems through the consequences of climate variability and climate change, which is usually manifested through in erratic and uncertain rainfall distribution. Rainfall is one of the climatic variables that largely determine the occurrence of drought and also influences the growth and development of vegetation. The analysis of the 36 years (1980-2016) annual RF showed a coefficient of variation ranging from 75.6%-202.7%, 70.8%-139.0% and 70.4%-157.2% over Fedis, Deder and Shinile Woredas respectively. The trends of annual and monthly total rainfall of the period 1980 to 2016 showed positive trends for Fedis and decreasing trends for Deder and Shinile. The highest annual RF was recorded in Deder compare to other study areas. The maximum total RF recorded in the month of April and August during Belg season (Feb-May) and Kiremt season (Jun-Sep) respectively over the study areas. The spatial distribution of drought frequency and its climate trend as well as different drought grades were obtained by employing the inverse distance weighting (IDW) interpolation method in ArcGIS 10.3.1 platform. So, the annual high drought frequency mainly concentrated in the Eastern some pocket areas of Fedis. From South to West part of Deder were observed slight to moderate drought. During Belg season, spatial pattern in SPI index, from central to South and Western part of Deder areas were observed slight to moderate drought respectively. In Kiremt season, spatial pattern in SPI index, Western part of areas was observed slight drought and in the Southern parts of areas was observed moderate drought. The annual drought frequency expanded from North to central part of Shinile woreda, large portion of area covered by moderate drought (was shown dry). In Shinile woreda, for some very small pocket areas were hit by slight and moderate MD while the majority of the areas under the influence of severe and very severe level of MD. During Belg (February-May) and Kiremt (June) the Meteorological drought risk map has been prepared by integrating from the drought index, SPI. The probabilities of occurrence of MD ranges were from 3 times to 13 times for no drought to very severe drought severity level. In Shinile woreda, during Belg season, in the tip of North small area is categorized very severe drought (11-13); in Kiremt season, from North Western to North tip areas were categorized moderate drought (5-6) to very severe drought (11-13). In Deder woreda, in Belg season, from South Eastern to South western was shown severe drought (7-10). And South tip part of areas was shown very severe drought (11-13). Whereas, during Kiremt season, in North West parts of area was observed severe drought. In Fedis woreda, in Belg season, in East some part of pocket area was severe drought (11-13), whereas, during Kiremt season, in South tip and in western small parts of areas were observed severe drought (11-13). So, Meteorological drought risk can be viewed as a product of both exposures to the climate hazards and the vulnerability of farming or cropping practices to drought conditions. Thus, Meteorological drought can be useful to guide decision making process in drought monitoring and to reduce the impact of drought on agricultural production and productivity, while identifying appropriate sites for specific adaptation and mitigation actions. In general, this information is in agreement with the result obtained from the analysis climate and yields data to reduce the impact of drought on agricultural production in study areas.

Conclusion

There have been many climate related disasters, such as drought, that have led to famine in Ethiopia. The experience has led to the evolution of an excellent early warning and response mechanisms in Ethiopia. The NMSA provides weather forecasts and the DPPB provides the famine early warning system and assesses the food supply in the country. The situation of the 1980 to 2016 case teaches us that drought studies are very important for study areas because these areas are aid dependent and recurrent drought experienced. Drought early warning based on onset, secession, LGP and dry spell information have to be practiced by study areas. So with credible drought information with a longer lead time, the national government can and in western small parts of areas were observed severe drought (11-13). So, Meteorological drought risk can be viewed as a product of both exposures to the climate hazards and the vulnerability of farming or cropping practices to drought conditions. Thus, Meteorological drought can be useful to guide decision making process in drought monitoring and to reduce the impact of drought on agricultural production and productivity, while identifying appropriate sites for specific adaptation and mitigation actions. In general, this information is in agreement with the result obtained from the analysis climate and yields data to reduce the impact of drought on agricultural production in study areas indicate that the occurrences of drought in study areas showed temporal variation.

Recommendation

Based on the findings of the study, the following recommendations are suggested:

- Since meteorological drought severity levels vary spatially, so, for further study the selection of agricultural technologies and information (drought tolerance crops, the type of crop variety, other weather parameters and soil moisture conservation practices) should be made to fit in to the MD severity levels
- The methods as well as results of MD assessment and risk mapping are believed to be highly important for decision makers and stakeholders, who have a stake in the study area. However, it is recommended that further studies can build up on this work by including real time seasonal drought forecasts as model parameter so that stakeholders can get early warning information that helps to take necessary adaptation and reduce the impact of MD
- Implementation of the results of the study areas (onset, Cessation, Dry spell and LDP) used to get early warning information for farmers in study regions in deciding on crop types to cultivate, to plan sowing dates, to harvest water at the time of dry spell and others and reduce the impact of MD on crops.
- The spatial distribution of seasonal crop moisture deficiency will be different if changes in cropping patterns occur. Therefore, if used on an operational basis for drought risk and vulnerability assessment, the crop-climate database will need to be updated periodically as the distribution of crops changes in response to economic factors, climate trends, and government programs

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