USING VARIOUS VISUALIZATION TECHNIQUES TO OUTLINE VULNERABILITY TO DROUGHT IN THE REPUBLIC OF MOLDOVA

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Abstract: - Using visualization techniques to outline vulnerability to drought in the Republic of Moldova. This paper is a demonstration of use of various visualization techniques in representing the results on vulnerability to drought at various times scales in the Republic of Moldova. Monthly precipitation totals from simulated data at 10 km horizontal resolution with a regional climatic model RegCM, CRU TS2.10 land observation data set at 0.5°x0.5° horizontal resolution and observations recorded at 15 meteorological stations in Moldova were used to compare the annual cycle of precipitation, seasonal variability of precipitation, and the spatial and temporal evolution of drought.

The Empirical Orthogonal Functions (EOF) analysis was used to identify the principal modes of seasonal variability of precipitation over the country and the Standardized Precipitation Index (SPI) was calculated as an indicator of drought or wetness at various time scales (1 to 24 months). The period of analysis was 1960-1997.

One of the visualization techniques was the Grid Analysis and Display System (GrADS) which is an easy manipulation and visualization tool for gridded and station meteorological data. We used it to represent the RegCM and CRU data on their grids. The kriging interpolation with a Gaussian model variance was selected as a suitable interpolation tool for the spatial distribution of the SPI and EOF over Moldova. For temporal diagram and mapping, Golden Software Surfer 9.0 and ArcGis Software 9.1 were used. The results presented with these visualization tools emphasize precipitation and drought characteristics over Moldova domain.

Key-Words: ArcGIS, Surfer, GrADS, drought, SPI, Republic of Moldova

1. Introduction

Drought is a recurring extreme climate event over land characterized by below-normal precipitation over a period of months to years. It is defined as a dry spell relative to its local normal condition. Drought is often assessed in of meteorological, agricultural, terms hydrological, and socio-economic impacts. Large areas of Europe including Republic of Moldova have been affected by drought durin g the 20th century. Recent severe and prolonged droughts have highlighted Europe's vulnerability to this and alerted natural hazard the public, governments, and operational agencies about many socio-economic problems accompanying water shortage and to the need for drought

mitigation measures.

The Republic of Moldova is among the eastern European countries affected by extreme drought. The temperature/ changes of precipitation ratio over the year shows that although Moldova's baseline climate only for the end of summer and the beginning of autumn were characterized as semiarid, it is likely that in the future there would be significantly longer and deeper dry spells. In particular, according to the results of six General Circulation Model experiments based on A2 and B2 SRES scenarios, Moldova will face warmer and wetter winters and hotter and drier summers and autumns. The projected annual decrease of precipitation in association with increase of

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temperature would likely stimulate strong humidity deficit inducing droughts (Corobov and Overcenco, 2007). Precipitation is the primary factor controlling the formation and persistence of drought conditions, but evapotranspiration is also an important variable.

In the present study, the Standardized Precipitation Index (SPI), originally developed by McKee et al. (1993) was adopted to assess the drought characteristics in the Republic of Moldova based on regional climate model (RegCM) simulations which were compared with station and CRU observation data sets.

is well recognized that It Global Circulation Models (GCMs) can reproduce reasonably well climate features on large scales (global and continental) but their accuracy decreases when proceeding from continental to regional and local scales because of the lack of resolution (Meehl et al., 2007). This is especially true for surface fields, such as precipitation and surface air temperature, which are critically affected by topography and land use. However, in many applications, particularly related to the assessment of climate-change impacts, the information on surface climate change at regional to local scale is fundamental.

One alternative to bridge the gap between the climate information provided by GCMs and that needed in impact studies is nesting of a fine scale limited area model (or Regional Climate Model, RCM) within the GCM. Such an approach have been used in the framework of the EU-project CECILIA (Central and Eastern Europe Climate Change Impact and Vulnerability Assessment). The regional climatic model ICTP RegCM3 centered over Romania and including the Republic of Moldavia was run at a horizontal resolution of 10 km for the current climate (19601-1990) and under SRES A1B scenario for 2021-2050 and 2071-2100 periods (Halenka, 2010, Boroneant et al., 2011).

In this paper RegCM3 simulations were compared with observations at 15 meteorological stations in Moldova and gridded CRU TS2.10 land observation data set in order to characterize seasonal precipitation variability and drought characteristics in terms of temporal evolution and spatial distribution over the Republic of Moldova. Appropriate manipulation and visualization techniques like Grid Analysis and Display System (GrADS), Golden Software Surfer8 and ArcGis Software 9.1 (ArcMap and ArcCatalog) were used for representing the results.

2. Data and Methods 2.1 Data

Monthly precipitation totals simulated with the Beta version of the regional climatic model ICTP_RegCM3 at a horizontal resolution of 10 km have been used. The ICTP_RegCM model was originally developed (Giorgi et al, 1993) and then augmented and used in various reference and scenario simulations (Giorgi et al., 1994a, 1994b; Pal et al., 2004).

The RegCM simulations conducted in **CECILIA-FP6** Project covered a domain (41.016°N-50.175°N; 14.095°E-36.192°E) centered over Romania (46°N, 25°E) (Boroneant et al, 2009; Boroneant et al, 2011; Halenka, 2010). The simulations were driven by ERA40 Reanalysis double nested from 25 km RegCM run for the period 1960-1997. For this study a domain centered over Republic of Moldova (45.01°N-49.01°N; 26.52°E-30.48°E) was considered [Fig. 1]. The domain contains 8010 grid points (90 lat x 89 lon). The topography of the RegCM represented with GrADS [Fig. 1a] and the real topography of the terrain and station distribution represented with ArcGIS [Fig1b] point out on the differences in altitude between the RegCM and reality. The observation data used to compare the results with RegCM simulations were the CRU TS2.10 data set

(http://www.cru.uea.ac.uk/cru/data/hrg/cru_ts_2 .10) and 15 station observations. The horizontal resolution of CRU TS2.10 land observation data set is 0.5° lat x 0.5° lon. The domain contains 100 grid points (10 lat x 10 lon). The period of analysis is 1960-1997.



Fig. 1 The topography of the RegCM represented with GrADS (a) and the real topography of the terrain and station distribution represented with ArcGIS (b).

2.2 Methods

2.2.1 EOF Analysis

To identify the principal modes of variability of seasonal precipitation over the territory of the Republic of Moldova the Empirical Orthogonal Functions (EOF) analysis (Preisendorfer 1988, von Storch 1995) has been performed. EOF techniques aims at finding a new set of variables that captures most of the observed variance from data through a linear combination of the original variables allowing such a convenient method for studying the spatial and temporal variability of the time series in a given domain. This method splits the temporal variance of data into orthogonal patterns called empirical eigenvectors. The EOF analysis was performed with the series of seasonal precipitation totals of gridded RegCM and CRU data sets and at 15 stations. The EOFs were calculated on detrended and standardized anomalies.

2.2.2 SPI

Drought was analyzed using the SPI (McKee et al. 1993, 1995) which has been considered the most reliable index for quantifying the intensity, duration and spatial

extent of drought on multiple time scales (Lloyd-Hughes and Saunders, 2002). It is an excellent tool for research on spatial analysis, since it removes the temporal effects of various precipitation magnitudes. The SPI is computed by fitting a probability density function to the frequency distribution of precipitation summed over the time scale of interest. Each probability density function is then transformed into a standardized normal distribution and the anomaly strength is classified into 7 categories [Table 1.]. The SPI was calculated at time scales from 1 to 24-months. At these time scales it is relevant to describe drought conditions for a range of meteorological, agricultural and hydrological applications and also socioeconomic impact.

2.2.3 Visualization techniques

In this paper three visualization techniques have been used to represent the results on seasonal precipitation variability and vulnerability to drought at various times scales in the Republic of Moldova. The Grid Analysis and Display System (GrADS) is an easy manipulation and visualization tool for gridded and station meteorological data (http://www.iges.org/grads/). It was used to represent the RegCM simulations and results on gridded data.

Geostatistical methods allow the spatial meteorological interpolation of data in relationship with topographic attributes. The basic tool of geostatistics and kriging is the semivariogram, which describes the spatial dependence between the estimation points and the measurement points in a function of distance. Kriging is a very flexible geostatistical gridding method. This method produces maps from irregularly spaced data. Surfer includes two kriging types: point and block kriging. Both point kriging and block kriging generate an interpolated grid. Block kriging estimates the average value of the rectangular blocks centred on the grid nodes. Since block kriging is estimating the average value of block, it generates smoother contours. Furthermore, since block kriging is not estimating the value at a point, it is not a perfect interpolator. Point kriging estimates the values of the points at the grid nodes. Thereby, in this study ordinary point kriging has been applied. An important component of the kriging gridding method is the variogram. The ordinary kriging interpolation with a Gaussian variogram was selected as a suitable interpolation tool for the spatial distribution of the SPI and EOF over Moldova.

For 3D-diagram and contouring mapping, Golden Software 8.0 and ArcGis Software 9.1 have been used.

Class no	Category	SPI value	Probability%
1	Extremely dry	\le -2.00	2.3
2	Severely dry	-1.50 to -1.99	4.4
3	Moderately dry	-1.00 to -1.49	9.2
4	Near Normal	-0.99 to	68.2
		+0.99	
5	Moderately wet	+1.00 to	9.2
		+1.49	
6	Severely wet	+1.50 to	4.4
	~	+1.99	
7	Extremely wet	$\geq +2.00$	2.3

Tab. 1. Drought classification by SPI value and corresponding event probabilities

3. Results and Discussions 3.1 Model validation

The 30-year monthly means of precipitation totals simulated by RegCM forced by ERA40, CRU observations and station observations were calculated for the period 1961-1990 in each grid point of corresponding data set, and at station locations.



Fig. 2 Annual cycle of precipitation based on RegCM (ERA) simulations, CRU and observation datasets for the reference period 1961-1990.

Then, these means were spatially averaged and compared. The results show that the model systematically overestimates precipitation compared to CRU and station observations [Fig. 2]. The largest magnitudes of RegCM (ERA40) precipitation errors are observed in late spring (April, May) and summer months (June, July, August). This drawback is transferred to SPI which is based only on precipitation.

3.2 EOF analysis of seasonal precipitation

In order to identify the principal modes of variability of seasonal precipitation totals over Moldova domain, the EOF analysis was performed on RegCM simulated data, CRU and station observations. The explained variance of the EOF1 of seasonal precipitation totals ranges between 65 and 79.8% for RegCM simulated data, between 73.5 and 88.6% for CRU observations and between 56 and 85% for station observation, respectively [Table 2]. The least variance is explained during summer and the most variance is explained during winter for both CRU and stations observations and, during autumn for RegCM simulations. The explained variance of the EOF2 ranges between 7.2 and 12% for RegCM simulations and between 4.3 and 11% for observations. The least explained variance for the EOF2 is during winter for simulated precipitation and station observations while the most explained variance of EOF2 is shown during spring for simulated precipitation and during summer for both CRU and station observation data. The explained variance of the EOF3 ranges between 3.7 and 6.7% for RegCM simulations and between 3 and 8.6% for observations.

Because the patterns of the three leading EOFs do not differ essentially from season to season, only the loading patterns for winter precipitation totals are shown, as example, in Fig. 3. In the upper row are represented the leading EOFs of winter precipitation totals for station data, in the second row for CRU data and on the bottom row for RegCM forced by ERA40 simulated data. The patterns of EOF1 for the three data sets are presented in the left hand column of Fig. 3. The spatial coefficients have the same sign over Moldova domain, the simulated data showing lower values of the loading than the observations. The distribution over the country's territory of the spatial coefficients of the EOF2 roughly separates the eastern and western halves of country territory [Fig. 3 middle column] while for the EOF3 the southern most region has opposite sign of loading compared to the rest of the country [Fig. 3, right hand side column]. These spatial characteristics of the leading EOFs loadings point out on various driving factors for seasonal variability of precipitation totals in Republic of Moldova: large scale circulation, continentality, Black Sea and Mediterranean influences.

The EOF patterns of winter precipitation for both CRU and station data were produced with the Golden Software Surfer 8.0 while for simulated data with GrADS2.0.a9.

2	Explained variance (%)					
	wi	sp	su	au		
RegCM_ERA40						
EOF1	77.15	70.25	65.01	79.79		
EOF2	7.18	12.04	10.11	7.35		
EOF3	6.50	5.17	6.70	3.74		
CRU @ stations						
EOF1	88.62	81.15	73.50	86.90		
EOF2	5.28	6.96	11.13	4.87		
EOF3	3.13	5.89	6.50	4.64		
Station obs						
EOF1	84.98	75.48	57.95	84.86		
EOF2	4.30	6.82	9.83	5.03		
EOF3	3.39	3.33	8.63	3.07		

Tab. 2. The explained variance of the first three EOFs of seasonal precipitation

3.3 Drought persistence

Drought appears first at short time scales and if dry conditions persist, it develops at longer time scales. The use of several time scales of SPI takes into account the role of antecedent conditions in quantifying drought severity, allowing a better understanding of time scales of water supplies. The SPI was calculated in each grid point of Moldova domain and at each station for accumulated periods ranging from 1 to 24-months. Then, the resulted SPI at 8010 grid points for RegCM simulation, at 100 grid points for CRU observation, respectively, and at 15 stations were averaged to get an aggregated SPI at country level for each accumulated period. To aggregate all times

scales of SPI from 1 to 24 months into the diagrams of Fig. 4 the Golden Software Surfer 8.0 was used. Each diagram contains a large volume of information for SPI (12 months x 38 years x 24 SPI time scales).



Fig. 3 The spatial patterns of the first three EOFs of winter precipitation.



Fig. 4 Temporal evolution of SPI at time scales from 1 to 24 months for Moldova domain for observations at 15 stations (top), CRU observations (middle) and RegCM simulations (bottom).

The temporal evolution of dry and wet conditions for Moldova domain was represented [Fig. 4] at 15 station observations (top), CRU observations (middle) and RegCM simulations. Each diagram shows an aggregated picture of the evolution of moisture characteristics quantified by SPI pointing out on the development of drought from one month to 24 months. The persistence of drought at these time scales is relevant for a range of meteorological, agricultural and hydrological applications and also for socio-economic impact studies. The RegCM captures the main characteristics of droughts identified both in CRU and station observations. According to SPI values characterizing categories of dry and wet conditions [Table 1] the periods of drought in the Republic of Moldova were 1962-1965, 1972-1974, 1977-1978, 1985-1986, 1990-1992 and 1996 while the periods with wet conditions were 1966-1967, 1970-1971 and 1981-1982. All these periods were identified both in the simulations and observations. It is evident that drought conditions have noticeable increased in the country with) drought duration gaining in persistence during the period 1960-1997 (Potop, 2011). Fig. 4 also shows the evolution and persistence of an individual drought pointing out on the lack of precipitation which characterizes these periods and on its large scale driving factors. The Golden Software Surfer technique turned out to be the most appropriate tool to represent an integrated image of the evolution and persistence of dry and wet conditions based of SPI values calculated for 1 to 24 months accumulation periods over Republic of Moldova.

3.4 Temporal evolution

As the time scale for calculation the SPI increases from 1 to 24 months the wet and dry conditions can be clearly identified as well as their persistence. The antecedent conditions in SPI calculation for 12-month accumulation period point out on the persistence of dry and wet conditions for time lengths of some years. The temporal evolutions of the averaged SPI calculated at 12-month accumulation period for station (top) and CRU observations (middle)

and RegCM simulations (bottom) for Moldova domain are represented in Fig. 5. When compared with the observations, the model does quite well in identifying the major dry and wet periods and their persistence. The main dry periods identified in the simulations are 1961-1965, 1968-1969, 1974-1975, 1983-1984, 1986-1987, 1989-1990, 1993-1995. These periods are also identified in the observations with similar or more severe category of dryness. The wet periods identified in the simulation are 1966-1967, 1970-1973, 1977-1982, and 1996-1997. Most of these periods are also identified in the observations but the severity of wetness is higher in the simulation than in the observations due to the drawback of the RegCM model to generally overestimate precipitation [Fig. 2].

3.5 Case study – Severe drought of 1986

One of the most extensive and severe drought events during the last fifty years in the Republic of Moldova was recorded in 1986. It was characterized as an agricultural drought as it lasted for many months, mostly during the growing season, and it had a dramatic impact on various crops (Potop and Soukup, 2009, Potop, 2011). The spatial distribution of SPI calculated at 6-month accumulation period over the country territory for the months May and August was selected as case study to assess the regional climatic model RegCM ability to the vulnerability to outline drought in comparison with the observations. On the top panel of Fig. 6 are represented the spatial distribution of SPI at 6 month accumulation period for May 1986 and on the bottom panel for August 1986. The maps were constructed with the Golden Software Surfer8.0 based on SPI values at 15 stations (left), and on the closest grid point value to station coordinates for CRU observation (middle) and RegCM simulation (right).



Fig. 5 Temporal evolution of averaged SPI at 12 month accumulation period over Moldova domain at 15 stations (top), CRU grid points (middle), and RegCM grid points (bottom)

According to SPI06 values and its categories [Table 1] the drought of May 1986 based on station observations was characterized as severe in most of the regions and extreme in the eastern half of country. Based on CRU observations, the drought was severe and extreme, but the area with extreme drought is larger than that resulted from station observations. Compared to both observation data sets (station and CRU) it is obvious that the model can not capture the severity of drought [Fig. 6 top row]. This is because as shown above [Fig. 2] the model overestimates precipitation which is the main factor to characterize drought with SPI. Similarly, for August 1986 [Fig. 6 bottom row] based on station observation the drought was severe and extreme over the whole country, moderate in the southern half, severe and extreme in the northern half of the country based on CRU data, and

moderate in the north and the south and normal moisture conditions in the south-east of the country based on RegCM simulation. This result is again a consequence of model overestimation of precipitation. These results should be taken into consideration when the RegCM is used in assessing the current and projected vulnerability to drought in Republic of Moldova.

4. Summary and Conclusions

- Various visualization techniques have been used to characterize domain topography, seasonal precipitation variability, spatial and temporal evolution of drought for various accumulation periods ranging from 1 to 24 months.
- The visualization technique is a user choice and it depends on its availability and user preference.



Fig. 6 Spatial evolution of SPI at 6-month accumulation period during the driest growing season (April-September) of 1986 over Moldova domain at 15 stations (left), CRU grid points (middle), and RegCM grid points (right).

- Two observation data sets (15 stations and CRU data set) were used to compare the RegCM ability to reproduce the annual cycle of precipitation, seasonal variability of precipitation, spatial and temporal evolution of drought based on SPI calculated for 1 to 24 accumulation periods.
- The results show that the model overestimates precipitation and this drawback is transferred to SPI which turn out to show less intense droughts than the observations.
- However, the model capture the general characteristics of precipitation and drought in terms of spatial and temporal distribution and, with due uncertainty, it can be used for vulnerability to drought assessments for the Republic of Moldova.

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