Statistical Downscaling HadCM3 Model for Detection and Prediction of Seasonal Climatic Variations (Case Study: Khabr Rangeland, Kerman, Iran)

Amir Saadatfar\textsuperscript{a}, Hossein Barani\textsuperscript{b}, Abdol Reza Bahremand\textsuperscript{c}, Ali Reza Massah Bavani\textsuperscript{d}, Adel Sepehry\textsuperscript{e}, Ahmad Abedi Servestani\textsuperscript{f}

\textsuperscript{a}Ph.D Student, Dep. of Range Land Management, Agricultural Sciences & Natural Resources University of Gorgan. (Corresponding Author). Email: amir_saadat0045@yahoo.com.
\textsuperscript{b}Assistant Professor, Dep. of Range Land Management, Agricultural Sciences & Natural Resources University of Gorgan.
\textsuperscript{c}Assistant Professor, Dep. of Watershed and Arid Zone Management, Agricultural Sciences & Natural Resources University of Gorgan.
\textsuperscript{d}Assistant Professor, Irrigation Group, Abureyhan University, Tehran.
\textsuperscript{e}Professor, Dep. of Range Land Management, Agricultural Sciences & Natural Resources University of Gorgan.
\textsuperscript{f}Assistant Professor, Dep. of Agricultural Extension and Education, Agricultural Sciences & Natural Resources University of Gorgan.

Received on: 14/08/2013
Accepted on: 05/01/2014

Abstract. Rangelands are one of the most vulnerable parts concerning the climate changes’ impacts. These impacts are even stronger in the arid and semi-arid areas where rangeland ecosystems are in critical conditions. Therefore, it is crucial to figure out the actual dynamics of climate variations on the rangelands. The aim of this research was to determine climate changes in Khabr rangeland, Kerman, Iran. So, four meteorological data sets of HadCM3 model including minimum and maximum temperature, precipitation and radiation rates were used to assess climate changes in the region. In this regard, climate changes during 2011-2039 were assessed by downscaling HadCM3 data using LARS-WG model under three scenarios of B1, A2 and A1B. The results have showed that the rainfall rates of spring and summer would have declining trends under all three scenarios. Minimum and maximum temperature rates would increase in all seasons, and just radiation one showed a decreasing trend for winter. Based on A1B scenario, minimum and maximum temperature rates had the highest increasing trend. Radiation and precipitation had the highest increasing and declining trends in the A2 scenario, respectively. Moreover, the increase in maximum and minimum temperature rates was averagely greater than the past and consequently despite the increasing trend in minimum and maximum temperature rates, the increase in the mean temperature rate during this period would be expected. According to the results, Khabr rangeland’s climatic conditions will be significantly different in the next 30 years and long-term and strategic planning is necessary in consistent with the management policies with these conditions.

Key words: Climatic-Parameters, HadCM3, LARS-WG, Khabr rangelands
1. Introduction

Climatic changes are ascribed to human activities that change the structure of the global atmosphere and to natural climate variability observed over comparable time periods (IPCC, 2007). Human activities, particularly the burning of fossil fuels and changes in land uses are today believed to increase the atmospheric accumulation of greenhouse gases. This change of energy balance has led to make the atmosphere warm and consequently, climatic changes. Some researches show that climatic parameters such as mean annual global surface temperature have increased by about 0.3 - 0.6°C since the late 19th century, and it is anticipated to increase by 1-3.5 °C over the next 100 years (Solomon et al., 2007). The most important impacts of climatic changes on rangelands will probably change both pasture productivity and forage quality. However, there are other impacts on rangelands that managers will confront including botanical changes in vegetation composition, pests, diseases, soil erosion, animal husbandry and health (Hall et al., 1998). Currently, global climate models are the only reliable tools available for simulating the response of the universal climate system to the increase of greenhouse-gas concentrations (IPCC-TGIA, 2007). Therefore, the fourth assessment report of AR4 was constructed according to a huge dataset about forthcoming climate-change projects by 18 worldwide groups, and climate experiences were run in several Global Climate Models (GCMs), and different scenarios (Semenov, 2008; Semenov and Stratonovitch, 2010). Scientists emphasize the needs for appropriate model testing against the observed data and the evaluation of uncertainty in future projections. Furthermore, climate models functions at spatial scales are considerably larger than those in which managerial decisions are made on the desired ecosystems.

Downscaling allows data obtained from global and regional models to be estimated for finer spatial scales. There is therefore a need to utilize the existing methods or develop new ones as appropriate ones from downscale climate model estimates to scales that are more relevant for site-specific decision-making. By the help of downscaling method, the GCM outputs can be correctly changed into surface variables in the scale of the basin under study. This method is based on the statistical model linking the climate simulated by the GCM, and the current climate characterized by instrumental data. This method has been widely applied to derive the daily and monthly precipitation rates at higher spatial resolutions for the impact assessments (Semenov and Barrow, 2002; Wilby et al., 2002).

LARS-WG stochastic weather generator simulates high-resolution temporal (daily) and spatial (site) climatic changes’ scenarios for a number of climate variables (e.g. precipitation, maximum/minimum temperature, and solar radiation rates). Scenarios have combined the changes with climatic variables such as duration of dry and wet spells or temperature variability derived from the daily output from GCMs (Semenov and Barrow, 2002). Future climate scenarios are stochastically generated by adjusting the parameters with respect to the directions proportion to the changes projected by a GCM. Further details of the LARS-WG can be found in the user manual (Semenov and Barrow, 2002). The main advantage of weather generators is their ability to generate multiple climate scenarios of daily climatic variables at a local station while making them very useful for the risk assessment studies. Therefore, they will be used within this study for the manufacturing of future climatic scenarios. The aim of this research was to determine climatic change in Khabr rangeland, Kerman, Iran. So, four
meteorological datasets of HadCM3 model including minimum and maximum temperature, precipitation and radiation rates were used to assess climatic change in the region.

2. Materials and Methods

The field research is a part of Khabr National Park’s rangelands located in Kerman Province in the south-east of Iran (29° 14’ N, 56° 35’ W, see Fig. 1). According to Emberger method, the region climate is arid frigid. The study site receives about 261 mm of annual precipitation that mostly occurs during November and May. The annual mean temperature is 17.6 °C. According to gaussen ambrothermic diagram, the region aridity period is 7 months. Low precipitation and high vapour-transpiration in this area have led to a severe drought in the recent decades.

Fig. 1. Location of case study, Khabr- Baft

Fig. 2. Research method diagram
2.1. Case study  
In this study, four main sources of data were used which are as follow:  
1- Daily historical temperature and precipitation data of Baft meteorological station during 1898-2011 (minimum temperature (T-min), maximum temperature (T-max), precipitation (P) and sunshine hours).  
2- Daily projected data of HadCM3 during 2011-2039 (T-min, T-max and P and sunshiny hours) that were resulted from GCM-runs for the Third Assessment Report (TAR) based on the IPCC-SRES scenario of A2.  
3- Monthly projected data of HadCM3 during 2011-2039 (T-min, T-max, P and sunny hours) based on the scenario of B1.  
4- Monthly projected data of HadCM3 during 2011-2039 (T-min, T-max, P and sunny hours) based on the scenario of A1B.  

As it has been shown in the research method diagram (see Fig. 2), LARS-WG model was applied to simulate climatic parameters. This model is one of the weather generators which can be utilized for the simulation of weather data at a single site (Rasco et al., 1991) under current and future climatic conditions. These data are in the shape of daily time-series for a set of climatic variables namely, precipitation (mm), maximum and minimum temperature and solar radiation rates (MJm-2day-1).  

LARS-WG is based on the series weather generator described by Rasco et al. (1991). It used semi-empirical distributions for the lengths of wet and dry day series, daily precipitation and daily solar radiation. Suitable large-scale predictors are selected using correlation analyses and partial correlations between predictors (large-scale atmospheric variables) and predictants (local surface variables) in the area under study.  

The simulation of a precipitation event is modelled as an alternate wet and dry series where a wet day is defined to be a day with the precipitation rate of > 0.0 mm. The length of each series is randomly selected from the wet or dry semi-empirical distributions for the month when the series starts. In determining the distributions, the observed series are also allocated to the month when they start. For a wet day, the precipitation value is generated from the semi-empirical precipitation distribution for the particular month independent from the length of the wet series or the amount of precipitation in previous days. Daily minimum and maximum temperature rates are considered as stochastic processes with daily means and daily standard deviations in the wet or dry status of the day. The technique used to simulate the process is very similar to that one presented by Rasco et al. (1991). LARS-WG performs the process of generating synthetic weather data with three distinct steps:  
1. Model Calibration - SITE ANALYSIS – the observed weather data are analysed to determine their statistical characteristics. This information is stored in two parameter files.  
2. Model Validation - QTEST - the statistical characteristics of the observed and synthetic weather data are analysed to determine whether there are any significant differences statistically.  
3. Generation of Synthetic Weather Data - GENERATOR - the parameter files derived from the observed weather data during the model calibration process are used to generate synthetic weather data having the same statistical characteristics as the original observed data, but they are different on a day-to-day basis. Synthetic data corresponding to a particular climatic change scenario may also be generated by applying global climate model derived from the changes in precipitation, temperature and solar radiation rates for the LARS-WG parameter files.  

At the beginning, the processing and sorting of climatic parameters were
performed. Then, the model was run for the base period and model calibration was completed. The next step was evaluated using statistical methods such as NASH (Eq. 1), Root mean square error (Eq. 2), Mean absolute error (Eq. 3) and Bias model performance (Eq. 4).

\[
\text{NA} = 1 - \frac{\sum_{i=1}^{n} (Y_i - X_i)^2}{\sum_{i=1}^{n} (X_i - \bar{X})^2}
\]

\[
\text{RSME} = \sqrt{\frac{\sum_{i=1}^{n} (X_i - Y_i)^2}{n}}
\]

\[
\text{MAE} = \frac{\sum_{i=1}^{n} |X_i - Y_i|^2}{n}
\]

\[
\text{Bias} = \frac{1}{n} \sum_{i=1}^{n} (X_i - Y_i)^2
\]

Where

\(X_i\) = observed data

\(Y_i\) = Simulated data

n = Total number of samples

After reviewing the results of the statistical methods, high performance of Lars-WG Model was demonstrated. With this model, daily value of meteorological parameters during 2011-2039 was simulated. The simulation was done under A2, A1B and B1 scenarios. Monthly normal climatic parameters were calculated for 2011-2039 based on three scenarios. Then, monthly and seasonal variations of these parameters were obtained under the scenarios of A2, A1B and B1.

3. Results

Comparison of results showed no significant difference between the modelled values and the actual values (P<0.05). (Table 1), shows Pearson correlation coefficient of all parameters. These values represent the accuracy of LARS-WG to simulate the climatic parameters. In (Table 2), NASH (NA), Root Means Square Error (RMSE), Mean Absolute Error (MAE) and Bias show high accuracy (when the value of NA is closer to one, Bias is closer to zero, RMSE and MAE have the minimum values whereas the observed and simulated values have more similarities). Based on these results during 1989-2010, the observed and simulated values have high similarities. Observation and modelling values of parameters such as the absolute minimum temperature, absolute maximum temperature, precipitation and sunshine hours with a standard deviation coefficient show that this model can simulate the trends within the data very well.

<table>
<thead>
<tr>
<th>BAFT Station</th>
<th>Rainfall</th>
<th>Min Tem</th>
<th>Max Tem</th>
<th>Sunshine Hours</th>
</tr>
</thead>
<tbody>
<tr>
<td>Correlation</td>
<td>0.98</td>
<td>0.99</td>
<td>0.99</td>
<td>0.98</td>
</tr>
<tr>
<td>Weighted correlation</td>
<td>0.87</td>
<td>0.98</td>
<td>0.98</td>
<td>0.96</td>
</tr>
</tbody>
</table>

**=Significant at 1% probability level

<table>
<thead>
<tr>
<th>BAFT Station</th>
<th>Rainfall</th>
<th>Min Tem</th>
<th>Max Tem</th>
<th>Sunshine Hours</th>
</tr>
</thead>
<tbody>
<tr>
<td>MAE</td>
<td>4.14</td>
<td>0.15</td>
<td>-0.02</td>
<td>-0.1</td>
</tr>
<tr>
<td>RSME</td>
<td>5.26</td>
<td>0.29</td>
<td>0.22</td>
<td>0.2</td>
</tr>
<tr>
<td>Bias</td>
<td>2.41</td>
<td>0.007</td>
<td>-0.07</td>
<td>-0.08</td>
</tr>
<tr>
<td>NA</td>
<td>0.95</td>
<td>0.99</td>
<td>0.99</td>
<td>0.96</td>
</tr>
</tbody>
</table>

The observed and simulated climatic parameters during 1989-2010 by LARS-WG model are shown in (Figs. 3-6). It represents the ability of the model to generate daily data on climate parameters which may be consistent with the research conducted by Semenov (2007).
Mean monthly values of climatic parameters and their changes under the baseline period and future scenarios (HadCM3 model scenarios) are given in Table 3.

Table 3. Mean monthly values of climatic parameters and their changes under the baseline period and future scenarios

<table>
<thead>
<tr>
<th>Parameters</th>
<th>Scenarios</th>
<th>Jan</th>
<th>Feb</th>
<th>Mar</th>
<th>Apr</th>
<th>May</th>
<th>Jun</th>
<th>Jul</th>
<th>Aug</th>
<th>Sep</th>
<th>Oct</th>
<th>Nov</th>
<th>Dec</th>
</tr>
</thead>
<tbody>
<tr>
<td>Normal Rainfall</td>
<td>Baseline</td>
<td>55</td>
<td>48.3</td>
<td>41.9</td>
<td>22.2</td>
<td>8.5</td>
<td>2</td>
<td>5.5</td>
<td>9.2</td>
<td>0.9</td>
<td>3.3</td>
<td>6.8</td>
<td>54.7</td>
</tr>
<tr>
<td></td>
<td>A1B</td>
<td>66.8</td>
<td>45</td>
<td>38.2</td>
<td>18.5</td>
<td>9.3</td>
<td>1.5</td>
<td>5.6</td>
<td>11.0</td>
<td>0.9</td>
<td>3.4</td>
<td>6.8</td>
<td>39.7</td>
</tr>
<tr>
<td></td>
<td>A2</td>
<td>66.6</td>
<td>45</td>
<td>38.2</td>
<td>18.7</td>
<td>9.6</td>
<td>1.4</td>
<td>5.1</td>
<td>10.1</td>
<td>0.9</td>
<td>3.38</td>
<td>6.5</td>
<td>39.2</td>
</tr>
<tr>
<td></td>
<td>B1</td>
<td>65.7</td>
<td>45.2</td>
<td>39.6</td>
<td>19.6</td>
<td>10.0</td>
<td>1.5</td>
<td>5.5</td>
<td>11.0</td>
<td>0.9</td>
<td>3.5</td>
<td>6.7</td>
<td>39.2</td>
</tr>
<tr>
<td>Variations</td>
<td>A1B</td>
<td>14.8</td>
<td>-3.3</td>
<td>-3.7</td>
<td>-3.7</td>
<td>0.8</td>
<td>-0.6</td>
<td>0.1</td>
<td>1.9</td>
<td>0</td>
<td>0.1</td>
<td>0</td>
<td>-6</td>
</tr>
<tr>
<td></td>
<td>A2</td>
<td>14.6</td>
<td>-2.8</td>
<td>-3.7</td>
<td>-3.6</td>
<td>1.1</td>
<td>-0.6</td>
<td>-0.3</td>
<td>0.9</td>
<td>0</td>
<td>0.08</td>
<td>-0.31</td>
<td>-6.6</td>
</tr>
<tr>
<td></td>
<td>B1</td>
<td>13.7</td>
<td>-3.1</td>
<td>-2.4</td>
<td>-2.6</td>
<td>1.5</td>
<td>-0.6</td>
<td>0</td>
<td>1.8</td>
<td>0</td>
<td>0.2</td>
<td>-0.1</td>
<td>-6.7</td>
</tr>
<tr>
<td>Minimum temperatures</td>
<td>Baseline</td>
<td>-2.3</td>
<td>-0.4</td>
<td>3.0</td>
<td>7.8</td>
<td>13</td>
<td>17.3</td>
<td>19.5</td>
<td>18.2</td>
<td>14.8</td>
<td>9.5</td>
<td>3.7</td>
<td>-0.04</td>
</tr>
<tr>
<td></td>
<td>A1B</td>
<td>-1.0</td>
<td>0.2</td>
<td>4.5</td>
<td>8.5</td>
<td>13.8</td>
<td>17.6</td>
<td>19.9</td>
<td>18.8</td>
<td>15.5</td>
<td>10</td>
<td>5</td>
<td>1.3</td>
</tr>
<tr>
<td></td>
<td>A2</td>
<td>-1.2</td>
<td>-0.2</td>
<td>4.0</td>
<td>8.1</td>
<td>13.6</td>
<td>17.7</td>
<td>19.9</td>
<td>18.7</td>
<td>15.6</td>
<td>9.9</td>
<td>4.6</td>
<td>0.9</td>
</tr>
<tr>
<td></td>
<td>B1</td>
<td>-1.2</td>
<td>0.2</td>
<td>4.8</td>
<td>8.6</td>
<td>13.8</td>
<td>17.6</td>
<td>19.8</td>
<td>18.5</td>
<td>15.3</td>
<td>10</td>
<td>5</td>
<td>1.2</td>
</tr>
<tr>
<td>Variations</td>
<td>A1B</td>
<td>1.3</td>
<td>0.6</td>
<td>1.4</td>
<td>0.7</td>
<td>0.8</td>
<td>0.3</td>
<td>0.4</td>
<td>0.5</td>
<td>0.7</td>
<td>0.6</td>
<td>1.3</td>
<td>1.3</td>
</tr>
<tr>
<td></td>
<td>A2</td>
<td>1.1</td>
<td>0.2</td>
<td>0.3</td>
<td>0.6</td>
<td>0.4</td>
<td>0.4</td>
<td>0.5</td>
<td>0.8</td>
<td>0.4</td>
<td>0.9</td>
<td>0.95</td>
<td>0.95</td>
</tr>
<tr>
<td></td>
<td>B1</td>
<td>1.1</td>
<td>0.6</td>
<td>1.7</td>
<td>0.8</td>
<td>0.8</td>
<td>0.3</td>
<td>0.3</td>
<td>0.6</td>
<td>0.6</td>
<td>1.3</td>
<td>1.2</td>
<td>1.2</td>
</tr>
<tr>
<td>Maximum temperatures</td>
<td>Baseline</td>
<td>8.7</td>
<td>11.1</td>
<td>15</td>
<td>20.3</td>
<td>26.3</td>
<td>30.8</td>
<td>32.4</td>
<td>31.3</td>
<td>28.1</td>
<td>22.9</td>
<td>16.7</td>
<td>12</td>
</tr>
<tr>
<td></td>
<td>A1B</td>
<td>9.7</td>
<td>11.5</td>
<td>16.1</td>
<td>20.9</td>
<td>27.2</td>
<td>31.1</td>
<td>32.8</td>
<td>31.8</td>
<td>28.9</td>
<td>23.6</td>
<td>17.8</td>
<td>13</td>
</tr>
<tr>
<td></td>
<td>A2</td>
<td>9.5</td>
<td>11.1</td>
<td>15.6</td>
<td>20.5</td>
<td>27</td>
<td>31.1</td>
<td>32.8</td>
<td>31.8</td>
<td>28.9</td>
<td>23.5</td>
<td>17.5</td>
<td>12.7</td>
</tr>
<tr>
<td></td>
<td>B1</td>
<td>9.6</td>
<td>11.6</td>
<td>16.3</td>
<td>21</td>
<td>27.1</td>
<td>31</td>
<td>32.7</td>
<td>31.6</td>
<td>28.7</td>
<td>23.7</td>
<td>17.9</td>
<td>12.9</td>
</tr>
<tr>
<td>Variations</td>
<td>A1B</td>
<td>1.0</td>
<td>0.5</td>
<td>1.0</td>
<td>0.6</td>
<td>0.9</td>
<td>0.3</td>
<td>0.4</td>
<td>0.5</td>
<td>0.7</td>
<td>0.8</td>
<td>1.1</td>
<td>1</td>
</tr>
<tr>
<td></td>
<td>A2</td>
<td>0.8</td>
<td>0</td>
<td>0.6</td>
<td>0.2</td>
<td>0.7</td>
<td>0.3</td>
<td>0.4</td>
<td>0.5</td>
<td>0.8</td>
<td>0.6</td>
<td>0.8</td>
<td>0.7</td>
</tr>
<tr>
<td></td>
<td>B1</td>
<td>0.8</td>
<td>0.5</td>
<td>1.3</td>
<td>0.7</td>
<td>0.9</td>
<td>0.3</td>
<td>0.3</td>
<td>0.3</td>
<td>0.6</td>
<td>0.8</td>
<td>1.1</td>
<td>0.9</td>
</tr>
</tbody>
</table>
The results from the analysis of climatic parameters showed that in the study area, the most significant changes of precipitation during 2011-2039 computed as 6.7% reduction will occur under the A2 scenario. After that, A1B and B1 scenarios with the decrease rates of precipitation as 5.8% and 5.1% have the highest percentage of variations, respectively (Fig. 7). Mean changes in sunshine hours on an average of 0.04 hours will decrease under the B1 scenario and this parameter will increase under the A1B and A2 scenarios (Fig. 8).

The mean minimum and maximum temperature variations during 2011-2039 are increased under all three scenarios. The highest variation rate in these parameters will occur under the A1B scenario with an average of 0.82 and 0.73 °C (Figs. 9 and 10).

Analysis of seasonal parameters under climate scenarios for 2011-2039 showed that spring and summer rainfall variations under all three scenarios will decrease and the increase of them will occur only in winter (Fig. 11).
Generally, the minimum and maximum temperature rates will increase in all seasons regarding three scenarios. The A1B and A2 scenarios have the highest and lowest variations concerning the three scenarios, respectively (Figs. 12 and 13).

Data analysis showed that sunshine hours will have a decreasing trend in winter under all three scenarios and an increasing trend in the sunshine hours in the other seasons is shown in Fig. 14.

**4. Discussion**

In this paper, for evaluating and understanding the climatic changes in the Khabr rangeland in Kerman province during 2011-2039, long-term predictions of an atmospheric general circulation model (HadCM3) were downscaled by LARS-WG model. The results showed that this model has a high efficiency in the simulation of daily climatic parameters which confirms the results of research conducted by Elashmay et al. (2005) and Semenov and Barrow (2002).

Downscaled data by HadCM3 model based on three scenarios of A1B, A2 and B1 for 2011-2039 have indicated that maximum and minimum temperature rates will increase with the average values of 0.61-0.82 and 0.53-0.73 (°C), respectively. Thus, the average temperature rate will be increased. Furthermore, an increasing trend will be observed in the number of sunshine hours. On the other hand, precipitation rate will be decreased to 12-14 (mm) under all three scenarios.

These results are in agreement with the results of Kousari et al. (2010) and Ragab and Prudhomme (2002).
Based on the results, the climate in the Khabr region for the next 30 years will markedly be different from the current situation as the largest decrease of precipitation rate (4-6%) will occur in spring and autumn as compared to the baseline precipitation in each scenario.

Minimum and maximum temperature rates also showed an increasing trend in spring and autumn in all three scenarios. This is consistent with the findings reported by Abasi et al. (2010). They stated that temperature rates would increase to 0.3°C, and most of it will be in winter. The observed trends in precipitation and temperature may cause snowfall to be reduced. In contrast, heavy rains will be increased. The consequence of that will be a reduction in the storage and supply of water through the gradual melting of snow in the mountainous areas. This heavy rainfall leads to increase the damage to natural ecosystem and people; on the other hand, fertile soils will be lost.

Undoubtedly, climatic changes are affecting dry lands and pastoral livelihoods. As a result, these areas will tend to become drier, and the existing water shortages will be worsened. In addition, climatic changes are likely to bring about even more erratic and unpredictable rainfall and most extreme weather conditions such as longer and more frequent droughts. Where this happens, the delicate balance on which pastoral systems depend is undermined.

Saadatfar et al. (2010) reported that changes in climatic parameters may cause to influence the growth behaviour of plants and rangeland exploitations. The changes in the number of freezing days may cause to alter the plant regeneration. Moreover, the earliest start of growing period leads to the changes of exploitation times as well as pastorals’ calendar. Apart from temporal changes, there are some impacts on spatial aspects in the mountainous ecosystems due to the mentioned thermal increase, i.e. changes in the existence boundaries of plant species and communities, their expansion to higher altitudes and shrinkage of meadows. Munanga et al. (2010) suggested that it is essential to amalgamate across information types (i.e. weather, climate, socio-economy, policy and ecology) to admonish those involved in decision-making better for the ecosystem management. The preparation of climate information and an understanding of ecosystem responses to climate changes and variability need to support all the planning and decision-making processes for the future.

With assessing the rangeland climate concerning the climatic changes’ scenarios, it is possible to realize the rates of changes in rainfall seasons and periods, severe droughts, floods, pest outbreaks, river flow decline, wetlands drying and expansion and intensification of dust storms in the future and based on them, risk in the rangelands should be effectively managed. This management approach can reduce the rangeland vulnerability and empower the rangeland beneficiaries in coping and adapting the critical situations.

**Literature**


پیش پیشینی تغییرات فصلی پارامترهای اقلیمی مرانگ با استفاده از ریز مقياس‌سنجی‌های داده‌های (مطالعه موردی: مرانگ خبر کرمان) مدل HadCM3

چیکده
مرانگ بکی از آسیب‌پذیرترین بخش‌ها، ناشی از اثرات تغییر اقلیمی می‌باشد. این تأثیر در مناطق خشک و نیمه‌خشک، در جنوب که مرانگ در شرایط محیطی قرار دارند، بیشتر است. بنابراین شناخت پویایی واقعی، تغییر اقلیمی بر مرانگ، بسیار مهم است. هدف از این تحقیق، تعیین تغییر اقلیم در منطقه خبر کرمان می‌باشد. در این راستا، چهار مجموعه از داده‌های هواشناسی مدل HadCM3 از داده‌های مدل HadCM3 که با استفاده از مدل LARS-WG در زمان 2039 - 2021 A1B و همچنین، می‌باشد که در حوزه دانشگاه علوم کشاورزی و منابع طبیعی قرار دارند. این رسیدگی به منظور ارزیابی تغییرات اقلیمی در دوره 2039 - 2021 A1B و A2 بود.

کلمات کلیدی: پارامترهای اقلیمی، HadCM3، LARS-WG