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Analyses of burned area of forest by adaptive neuro-fuzzy approach

ABSTRACT

Burned area of forest may be developed based on different factors. In this investigation the burned area of forest was analyzed by computational intelligence approach. The main goal was to analyze the influence of eight inputs on the burned area of forest. The method of ANFIS (adaptive neuro fuzzy inference system) was applied to the data. Eight inputs are considered: FFMC, DMC, DC, ISI, temp, RH, wind, rain. The ANFIS process was implemented to detect the dominant factors which affect the forecasting of the burned area of forest.

Keywords: ANFIS; forecasting; the burned area of forest.

1. INTRODUCTION

A number of studies have examined controls on high severity fire occurrence, but none have yet determined what controls the extent of high severity fire [1,2]. Global burned area algorithms provide valuable information for climate modellers since fire disturbance is responsible of a significant part of the emissions and their related impact on humans [3]. Forest fires have environmental impacts that create economic problems as well as ecological damage. Developing a means to predict the possible size of a fire shortly after it first breaks out has the potential to guide proper resource allocation for improved fire control and was the main motivation of this research [4]. Fire is an integral Earth system process, playing an important role in the distribution of terrestrial ecosystems and affecting the carbon cycle at the global scale [5]. Fire Weather Index (FWI) results in article [6] suggested the predicted increase in extreme summer heat events with global warming could increase burned area as firefighting resources are stretched beyond capacity. Managers of wildfire-prone landscapes in the Euro-Mediterranean region would greatly benefit from fire weather predictions a few months in advance, and particularly from the reliable prediction of extreme fire seasons [7].

A fire danger rating system should supply an objective answer to the question: 'What is the probability of a fire starting, spreading and doing damage today?' It enables fire managers to properly assess the levels of preparedness and the suppression resources needed to keep fire losses to a minimum. A fire danger rating system measures the variable elements which cause day to day changes in fire risk, and interprets the information gained. The information is used to:

- define the fire season
- determine appropriate fire prevention measures
- assess the likelihood of fire occurring
- determine fire suppression response and resources
- inform the public
- make decisions to close areas at high risk
- issue or cancel burn permits
- plan and conduct controlled burns

In this investigation adaptive neuro-fuzzy inference system (ANFIS) [8], was used to analyze burned area of forest based on three Fuel Moisture Codes and one Fire Behaviour Index

2. METHODOLOGY

The Fire Weather Index (FWI) System is the first part of the Canadian Forest Fire Danger Rating System (CFFDRS) introduced into New Zealand in 1980. It has proved to be a suitable fire danger rating system for this country. The FWI was evaluated for several seasons before it was

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introduced for the 1980-81 fire season. The FWI is based on weather readings taken at noon standard time and rates fire danger at the midafternoon peak from 2:00 - 4:00 pm. Weather readings required are:

- Air temperature (in the shade)
- Relative Humidity (in the shade)
- Wind speed (at 10 metres above ground level for an average over 10 minutes)
- Rainfall (For the previous 24 hours)

The Fire Weather Index has six components:

Three Fuel Moisture Codes

Fine Fuel Moisture Code (FFMC)

Duff Moisture Code (DMC)

Drought Code (DC)

Three Fire Behaviour Indices

Initial Spread index (ISI)

Build Up Index (BUI)

Fire Weather Index (FWI)

The main goal was to analyze the influence of FFMC, DMC, DC, ISI, temp, RH, wind, rain on the burned area of forest. Table 1 shows input parameters which are used in this investigation.

Table 1. Input parameters

Tabela 1. Ulazni parametri

Inputs	Parameters description
in1	Fine Fuel Moisture Code (FFMC)
in2	Duff Moisture Code (DMC)
in3	Drought Code (DC)
in4	Initial Spread index (ISI)
in5	temp
in6	RH
in7	wind
in8	rain

2.1. ANFIS methodology

Fuzzy inference system is employed of the ANFIS training and evaluation. In the process of identification of variables in the ANFIS architectures, the hybrid learning algorithms were applied. The functional signals progress until the 4th layer whereby the hybrid learning algorithm passes. Further, the consequent variables are found by the least squares estimation. In the

backward pass, the error rates circulate backwards and the premise variables are synchronized through the gradient decline order.

3. RESULTS

3.1. Evaluating accuracy indices

Forecasting performances of proposed model were presented as root means square error (RMSE), Coefficient of determination (R²) and Pearson coefficient (r). These statistics are defined as follows:

1) root-mean-square error (RMSE)

$$RMSE = \sqrt{\frac{\sum_{i=1}^n (P_i - O_i)^2}{n}} \tag{1}$$

2) Pearson correlation coefficient (r)

$$r = \frac{n \left(\sum_{i=1}^n O_i \cdot P_i \right) - \left(\sum_{i=1}^n O_i \right) \cdot \left(\sum_{i=1}^n P_i \right)}{\sqrt{\left(n \sum_{i=1}^n O_i^2 - \left(\sum_{i=1}^n O_i \right)^2 \right) \cdot \left(n \sum_{i=1}^n P_i^2 - \left(\sum_{i=1}^n P_i \right)^2 \right)}} \tag{2}$$

3) coefficient of determination (R²)

$$R^2 = \frac{\left[\sum_{i=1}^n (O_i - \bar{O}_i) \cdot (P_i - \bar{P}_i) \right]^2}{\sum_{i=1}^n (O_i - \bar{O}_i) \cdot \sum_{i=1}^n (P_i - \bar{P}_i)} \tag{3}$$

where P_i and O_i are known as the experimental and forecast values, respectively, and n is the total number of test data.

3.2. ANFIS results

A search was performed from the given inputs to choose the combinations of inputs (Table 1) which has the most impact on the burned area of forest. The parameters' impacts in the forecasting of the output was determined, as presented in Table 2. The input with the lowest training error (trn) has the most relevance to the burned area of forest. According the Table 2 the input parameter 6 has the highest influence on the burned area of forest forecasting since the input 6 has the smallest training error (trn). Table 2 shows also the numerical results for two and three parameters influence on the burned area of forest.

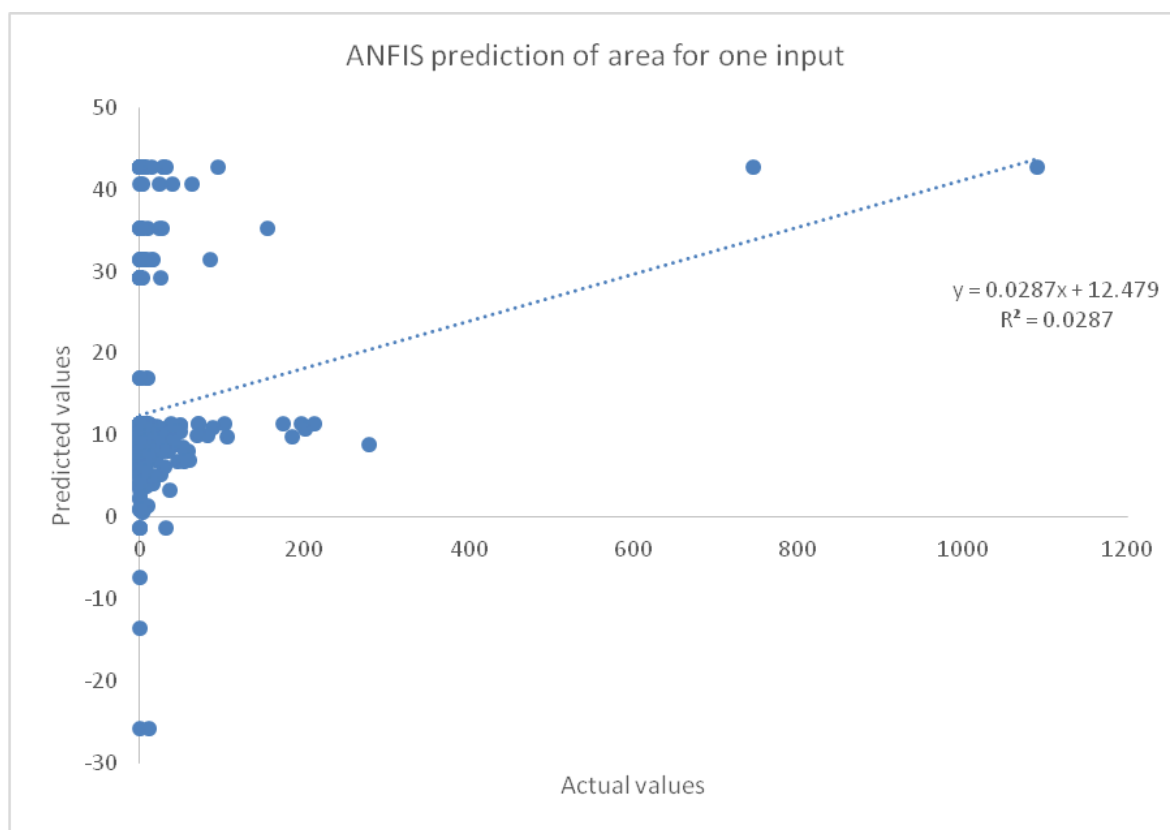
Table 2. Parameters influence on forecasting of the burned area of forest

Tabela 2. Uticaj parametara na predviđanje površine požara u šumi

in1 --> trn=70.8776, chk=55.1682	in1 in2 --> trn=70.5966, chk=80.0445	in1 in2 in3 --> trn=70.1852, chk=68.4009
in2 --> trn=70.7016, chk=55.5331	in1 in3 --> trn=70.6552, chk=58.2432	in1 in2 in4 --> trn=70.1788, chk=89.7701
in3 --> trn=70.7032, chk=55.4199	in1 in4 --> trn=70.7250, chk=55.4623	in1 in2 in5 --> trn=70.0338, chk=108.7148
in4 --> trn=70.8998, chk=55.1725	in1 in5 --> trn=70.4987, chk=65.6365	in1 in2 in6 --> trn=69.8834, chk=64.2655
in5 --> trn=70.6241, chk=54.7341	in1 in6 --> trn=70.4514, chk=55.7053	in1 in2 in7 --> trn=69.9907, chk=144.9729
in6 --> trn=70.5134, chk=55.4872	in1 in7 --> trn=70.5419, chk=63.0611	in1 in2 in8 --> trn=70.5862, chk=77.8195
in7 --> trn=70.6711, chk=55.4003	in1 in8 --> trn=70.8545, chk=55.2624	in1 in3 in4 --> trn=70.1868, chk=75.1055
in8 --> trn=70.9575, chk=59.5532	in2 in3 --> trn=70.2993, chk=56.4595	in1 in3 in5 --> trn=70.0067, chk=108.6739
	in2 in4 --> trn=70.4862, chk=56.3718	in1 in3 in6 --> trn=69.9181, chk=206.8819
	in2 in5 --> trn=70.2399, chk=56.0975	in1 in3 in7 --> trn=70.0876, chk=64.2259
	in2 in6 --> trn=70.1863, chk=55.5388	in1 in3 in8 --> trn=70.6360, chk=57.9246
	in2 in7 --> trn=70.2782, chk=55.9343	in1 in4 in5 --> trn=70.1054, chk=66.4705
	in2 in8 --> trn=70.6830, chk=69.3239	in1 in4 in6 --> trn=70.0034, chk=65.7524
	in3 in4 --> trn=70.4386, chk=55.8630	in1 in4 in7 --> trn=70.0656, chk=298.7084
	in3 in5 --> trn=70.4057, chk=55.2416	in1 in4 in8 --> trn=70.7082, chk=55.4288
	in3 in6 --> trn=70.1083, chk=55.4720	in1 in5 in6 --> trn=69.9719, chk=112.1698
	in3 in7 --> trn=70.2413, chk=56.0605	in1 in5 in7 --> trn=69.6507, chk=86.9506
	in3 in8 --> trn=70.6841, chk=55.4332	in1 in5 in8 --> trn=70.4951, chk=65.3891
	in4 in5 --> trn=70.4467, chk=55.2043	in1 in6 in7 --> trn=69.8960, chk=149.4115
	in4 in6 --> trn=70.3835, chk=55.5863	in1 in6 in8 --> trn=70.4470, chk=55.7860
	in4 in7 --> trn=70.4458, chk=55.7081	in1 in7 in8 --> trn=70.5214, chk=59.0786
	in4 in8 --> trn=70.8838, chk=59.5185	in2 in3 in4 --> trn=69.8389, chk=60.1344
	in5 in6 --> trn=70.1677, chk=55.6105	in2 in3 in5 --> trn=69.5835, chk=57.1709
	in5 in7 --> trn=70.0143, chk=55.0437	in2 in3 in6 --> trn=69.5317, chk=57.1178
	in5 in8 --> trn=70.6108, chk=54.7739	in2 in3 in7 --> trn=69.6546, chk=57.7443
	in6 in7 --> trn=70.0306, chk=55.7767	in2 in3 in8 --> trn=70.2910, chk=56.6503
	in6 in8 --> trn=70.5104, chk=56.1158	in2 in4 in5 --> trn=69.7771, chk=59.8863
	in7 in8 --> trn=70.6586, chk=55.3880	in2 in4 in6 --> trn=69.7242, chk=56.6838
		in2 in4 in7 --> trn=69.6987, chk=59.0423
		in2 in4 in8 --> trn=70.4842, chk=56.3771
		in2 in5 in6 --> trn=69.7615, chk=57.1683
		in2 in5 in7 --> trn=69.5104, chk=56.7480

		in2 in5 in8 --> trn=70.2345, chk=56.2462
		in2 in6 in7 --> trn=69.4327, chk=55.1262
		in2 in6 in8 --> trn=70.1859, chk=55.5362
		in2 in7 in8 --> trn=70.2524, chk=55.8709
		in3 in4 in5 --> trn=69.7581, chk=59.1198
		in3 in4 in6 --> trn=69.4794, chk=58.5765
		in3 in4 in7 --> trn=69.6766, chk=57.0705
		in3 in4 in8 --> trn=70.4288, chk=55.8391
		in3 in5 in6 --> trn=69.8140, chk=55.1576
		in3 in5 in7 --> trn=69.6071, chk=56.2251
		in3 in5 in8 --> trn=70.3833, chk=55.3118
		in3 in6 in7 --> trn=69.3915, chk=56.2117
		in3 in6 in8 --> trn=70.1058, chk=55.4898
		in3 in7 in8 --> trn=70.2179, chk=56.1407
		in4 in5 in6 --> trn=69.7383, chk=63.7406
		in4 in5 in7 --> trn=69.3746, chk=64.1600
		in4 in5 in8 --> trn=70.4373, chk=55.1902
		in4 in6 in7 --> trn=69.6787, chk=56.5632
		in4 in6 in8 --> trn=70.3693, chk=55.7489
		in4 in7 in8 --> trn=70.4333, chk=55.7438
		in5 in6 in7 --> trn=69.2943, chk=56.3454
		in5 in6 in8 --> trn=70.1663, chk=55.5662
		in5 in7 in8 --> trn=69.9925, chk=55.1766
		in6 in7 in8 --> trn=70.0279, chk=56.0244

Figure 1 shows ANFIS forecasting of the burned area of forest for the optimal combinations of the selected input parameters.



(a)

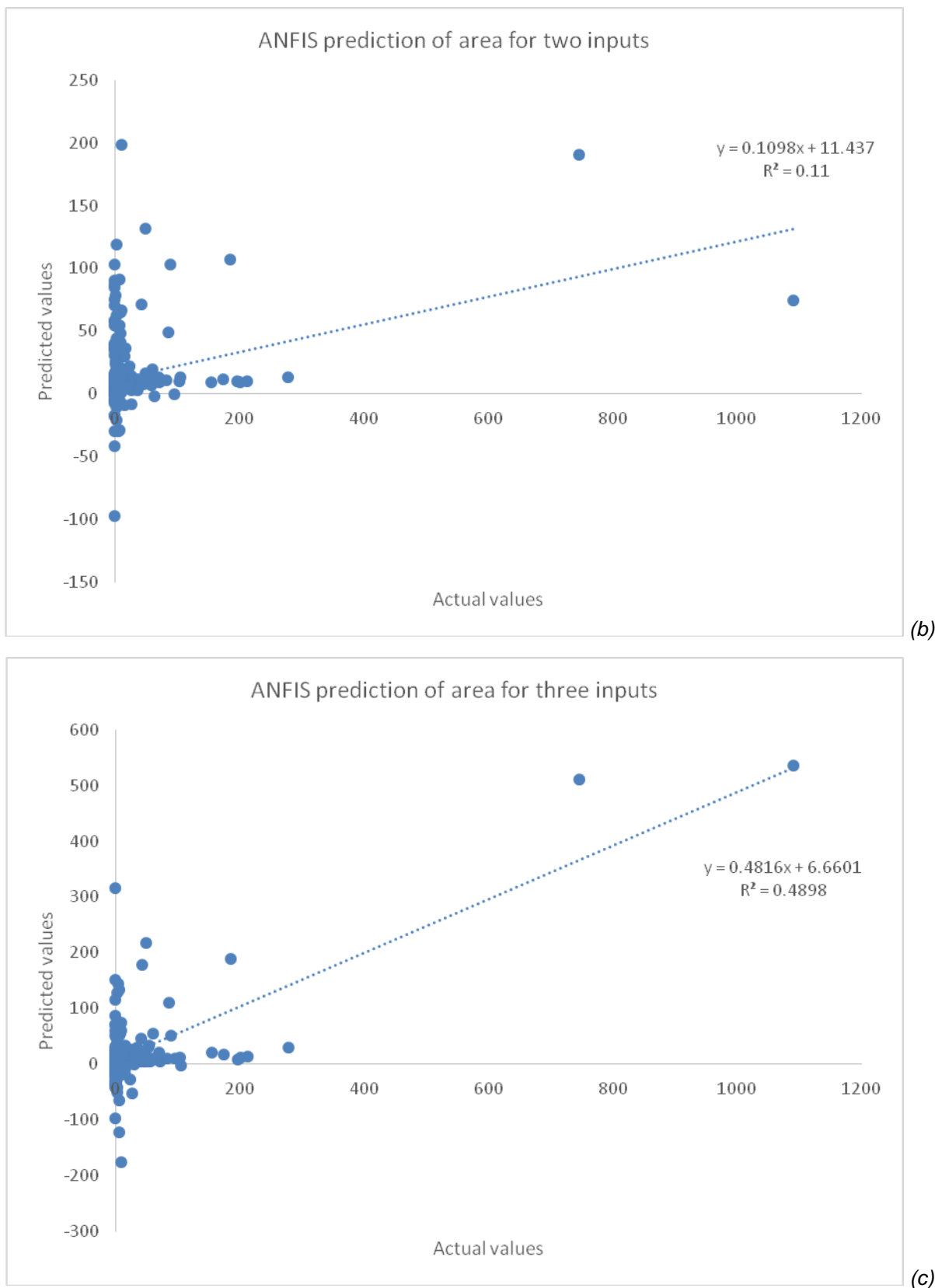


Figure 1: ANFIS plots for forecasting of burned area of forest for (a) one input, (b) two inputs (c) three inputs

Slika 1: ANFIS grafici predviđanja površine požara u šumi za (a) jedan ulaz, (b) dva ulaza (c) tri ulaza

Table 3 summarize the forecasting accuracy results for the selected inputs for the forecasting of the burned area of forest.

Table 3. Statistical results for forecasting of the burned area of forest for the three selected combinations

Tabela 3. Statistički rezultati predikcije površine požara u šumi

burned area of forest forecasting for one input	r	0.169341
	R ²	0.0287
	RMSE	62.67577
burned area of forest forecasting for two inputs	r	0.331658
	R ²	0.11
	RMSE	59.99479
burned area of forest forecasting for three inputs	r	0.699839
	R ²	0.4898
	RMSE	45.43143

4. CONCLUSION

Forecasting of the burned area of forest is complex task due to the many indicators which influence the burned area of forest. Therefore in this study was proposed an approach to overcome the forecasting difficulties of the burned area of forest by removing some parameters. The procedure is known as selection procedure and the main goal was to determine the parameters' influence on the output prediction.

IZVOD

ANALIZA POVRŠINE POŽARA U ŠUMI PRIMENOM NEURO FAZU TEHNIKE

Površina požara u šumi može da se analizira na bazi različitih ulaznih parametara kao i različitih kombinacija tih parametara. U ovom istraživanju je analizirana površina požara u šumi na osnovu različitih kombinacija ulaznih parametara. Glavni cilj je analizirati kako parametri utiču na površinu požara u šumi. ANFIS metodologija je korišćena u tu svrhu. Osam ulaznih parametara je korišćeno: FFMC, DMC, DC, ISI, temperatura, RH, vetar, kiša.

Ključne reči: ANFIS; predikcijač površina požara; šuma.

Naučni rad

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