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Prediction of visitors' thermal comfort in open urban areas

ABSTRACT

Thermal sensation of visitors could be important indicator for urban spaces based on physiological, psychological and behavioural conditions of the visitors. Therefore in this article thermal comfort of the visitors in open urban spaces was investigated. For such a purpose physical data and questionnaire study were used. In order to estimate the optimal conditions in the open urban areas there is need to perform prediction of the thermal comfort and thermal comfort in the areas. The prediction results could be used for optimal arrangement in the open urban areas based on thermal comfort of the visitors. Results of three methods were compares and analysed. The methods are based on artificial intelligence principle.

Keywords: Thermal comfort; Urban square; Neural network; Microclimatic.

1. INTRODUCTION

Rapid development of urban areas requires also acceptable thermal conform of visitors as well. There is need to investigate the thermal comfort based on the microclimatic environmental factors. There are different factors which could impact the thermal conform of the visitors. The factors could be social, cultural and recreational activities of the visitors. Thus there is need to create a methodology which will be based on all aspect of thermal comfort.

Psychological and sociocultural processes are important in assessment of thermal comfort [1]. One of the greatest environmental challenges for the sustainability of future cities is the mitigation of the urban heat island phenomenon and thus, improvement of outdoor comfort conditions for people [2]. There is need to bridge the gap between urban climatology and urban design in coastal temperate climate cities of developing countries [3]. Urban parks were measured to be thermally uncomfortable, but were perceived to be thermally comfortable [4]. Economic development of city could be improved by well-designed urban space could attract more visitors and to enhance their mood and behaviour and facilitate social interactions [5]. Also extreme weather conditions need to be counted during designing of open urban spaces since the weather could be vulnerable in the spaces [6]. Global warming factor needs to be included as well in the modelling of urban spaces [7].

Results in article [8] were shown there are relationship between indoor and outdoor thermal sensation. There are several investigations of the indoor thermal confirm but there are no similar investigations for open urban spaces [9-16]. Thermal comfort is investigated in urban canons and tropic zones [15-22].

In the study thermal comfort of visitors in open urban areas are investigated based on artificial intelligence methods [23-32]. The thermal comfort are analysed based on four parameters: physiologically predicted mean vote (PMV), the equivalent temperature (PET), the standard effective temperature (SET). Predictive models are created for the outputs with an advances algorithm based on artificial intelligence methods. As the input factors microclimatic parameters are used and questionnaire survey is performed of the visitors about their thermal comfort.

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J. Galjak et al.

2. METHODOLOGY

2.1. Experimental measurements of thermal comfort in open urban area

Urban locations in Serbia are analysed and corresponding data are acquired. The measurements are performed on the field for one whole year to cover all microclimatic and climatic conditions. The parameters which are measured are: air temperature (T_a) , globe temperature (T_g) , relative humidity (RH), globe radiation (R_s) , wind speed (W_s) and wind direction (W_{dir}) .

There are also questionnaire survey of the visitors on the open urban areas which is also conducted. The survey is based on demographic information, overall conform feeling, reasons to be there, visitors' behaviour. These input factors for the survey are: location, activity, age group, gender and locality.

Input factor "location" is based on the visitors' location in the urban area. There are four locations which are identified. The first location with open space with covering roofs. The second location is fully shaded open urban area. The third location is on the plants area. The fourth location is on the open urban space or urban square.

Input factor "activity" is based on visitors' activity on the open urban space. There are four activities in the study: sedentary, standing, walking, riding a motorcycle, exercising, riding a bicycle.

Input factor "age group" is based on the visitors' age as follows: child, teen, 18-24, 25-43, 45-54, 55-64 and more than 65.

Input factor "gender" has two numerical values based on the visitors' gender. Also the input factor "locality" has two numerical values for local visitors and for tourist visitors.

There are four outputs which are used in the study. These parameters represents the thermal comfort in the open urban area. The outputs are calculated by RayMan model [33].

2.2. Computational modelling by artificial intelligence methods

The first methodology is based on neural network (NN) principle. The acquired input/output data pairs are used for NN training and testing purpose. NN is capable to solve relationship of the highly nonlinear data like thermal comfort data. As training algorithm for the NN, extreme learning machine (ELM) [34, 35] is used. Although there are many different training algorithms for the NN, extreme learning machine us used since the ELM is faster than traditional training method like backpropagation algorithm. By using ELM more reliable generalization performances could be observed and training process is faster than backpropagation (BP) algorithm [36]. ELM training algorithm uses single feed forward neural networks. Basic principle of the ELM training procedure is determining of random input weights of the NN. In such a way training procedure is more relaxed and therefore there is need less time.

The predictive results of the NN with ELM training algorithm are then compared with computational methodology based on artificial intelligence like genetic programming (GP). Also as comparative results NN with ELM is compares with NN with BP training algorithm.

The predictive results of the applied methodologies are compared by three statistical indicators: mean square error (RMSE), the coefficient of determination (R2), and the Pearson correlation coefficient (r).

3. RESULTS

Figures 1-3 shows the predictive performances of the thermal comfort with the three methods based on the given inputs. Figure 2 shows the predictive results by NN with ELM for:

- Equivalent temperature (PET)
- Physiologically predicted mean vote (PMV)
- Standard effective temperature (SET)

One can note that the NN with ELM has the best predictive performances for the PET prediction according to the coefficient of determination. Based on the results in Figure 2 and 3 one can conclude that the NN with ELM has the better predictive performances for thermal comfort than NN with BP and GP results.

Table 1 shows the comparative results of the three methods based on the three statistical indicators. NN with ELM outperform the other two approaches based on the all statistical indicators. Also one can note that the NN with ELM has high prediction accuracy for the thermal comfort. In other words RMSE values should be smaller and R2 and r should be closer to 1 in order to be more precise. Based on R2 coefficient one can see that ELM approach has strongest correlation since R2 values are higher than 0.95.

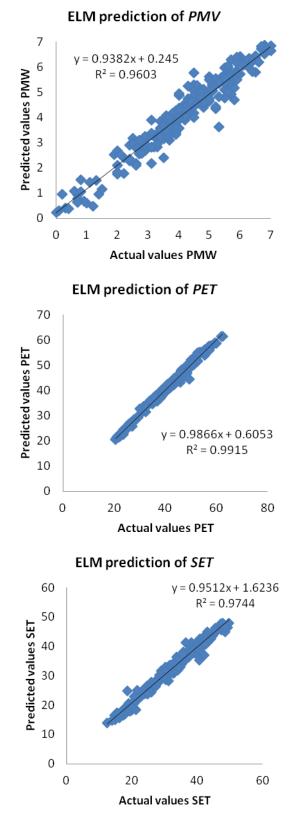
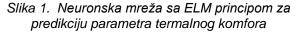


Figure 1. NN with ELM predictive results for thermal comfort parameters



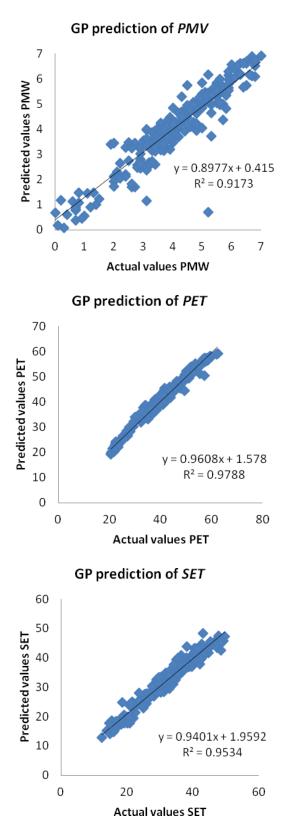
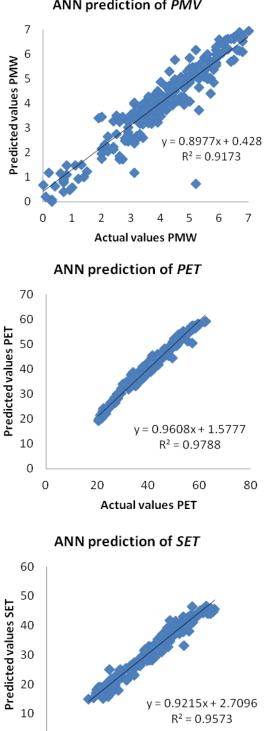


Figure 2. GP predictive results for thermal comfort parameters

Slika 2. GP metoda za predikciju parametra termalnog komfora



ANN prediction of PMV

1. Predictive performances of the three Table methods

Predicted values PMW	6 5 4 3 2 1 0	0	1	2 Actu	3 al vali	y = 4	R ² =	77x+0 0.917 6			
			ANN								
	70										
Predicted values PET	60										
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ıre	З.	NI	N with	BP pr	edict	ive r	esults	s for t	hermal		

Tabela 1. Prediktivne performanse tri metode

PMV	ELM	0.4001	0.9603	0.9799
FIVIV	ANN	0.5757	0.9173	0.9577
	GP	0.5751	0.9173	0.9577
PET	ELM	0.8754	0.9915	0.9957
FEI	ANN	1.3938	0.9788	0.9893
	GP	1.3938	0.9788	0.9893
SET	ELM	1.3107	0.9744	0.9871
351	ANN	1.7043	0.9573	0.9784
	GP	1.7534	0.9534	0.9764

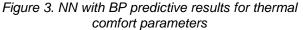
4. CONCLUSION

In the study a procedure is performed to make prediction of the thermal comfort of the visitors in open urban areas. The thermal comfort is presented with four output factors. Input factors are on microclimatic conditions based and a questionnaire survey. The visitors' responses are analysed based on their location, activity, gender, age and locality.

Computational intelligence approaches are used for the prediction of the thermal comfort. The approaches are based on neural networks and genetic programming principle. Neural network is trained with two training algorithm for comparative study. The training algorithm are extreme learning machine and back-propagation algorithm. Measurement is performed to acquire thermal conform input/output data pairs. The data is them used for the training and testing of the computational intelligence approaches. Based on prediction neural network with extreme learning machine shows the best predictive performances of the thermal comfort parameters.

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Slika 3. Neuronska mreža za BP principom za predikciju parametra termalnog komfora

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IZVOD

PREDIKCIJA TERMALNOG KOMFORA U OTVORENIM URBANIM POVRŠINAMA

Termalna osećajnost od strane posetilaca i turista je važan indicator za urbane površine na osnovi fizioloških, psiholoških i uslova ponašanja turista. Prema tome u ovom radu je analiziran termalni komfor posetilaca i turista u otvorednim urbanim površinama. Za tu svrhu su korišćeni fizički podaci kao anketa među turistima i posetiocima. Kako bi bili postignuti optimalni uslovi u otvorenim urbanism površinama potrebno je uraditi predikciju termalnog komfora u tim površinama. Rezultati poredikcije se mogu potom koristiti za optimalni aranžman u otvorenim urbanism površinama. Rezultati su dobijeni pomoću tri metode. Te metode su bazirane na osnovu veštačke inteligencije.

Ključne reči: termalni komfor; urbana površina; neuronska mreža; mikroklima

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