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# Prediction of wind blowing durations of Eastern Turkey with machine learning for integration of renewable energy and organic farming-stock raising

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ARTICLE INFO	ABSTRACT
Article History           Received         : 27/02/2019           Revised         : 12/06/2019	Applications which integrate wind energy and both agriculture and stock raising are increasingly becoming popular especially in Europe. Subject applications enable the land to be utilized in various favorable ways. In this study, by using a 5-year average wind data referring
Accepted : 17/09/2019 Available online : 30/09/2019	to Erzurum and Ardahan, two eastern cities of Turkey which are characterized by prevailingly an extensive cattle-raising, wind-blowing durations were calculated by Rayleigh distribution.
Keywords	Annual wind blowing durations for Erzurum and Ardahan ranged between 479.6-5825.7
Wind energy prediction	hours and 1643.6-6710.8 hours, respectively. The data obtained was predicted via artificial
Wind blowing duration	neural networks and output results indicate an prediction accuracy at 99% level thereupon.
Artificial neural networks	The integration of agricultural and stock raising activities with wind energy shall contribute
Machine learning	to environmental aspects as well increasing the efficiency and effectiveness in the region.
Organic farming, Livestock raising	

#### 1. INTRODUCTION

Global energy needs are increasing rapidly in accordance with the increasing population. Due to depletion and their environmental damage, the fossil fuels used today to meet the energy needs are being replaced by renewable energy sources. Wind energy is a renewable energy source that is increasingly used in our country. Wind energy generation in a certain region depends on the performance characteristics and operating conditions of the wind on that specific region. It would not be possible to determine a stable model for the system where and if the wind velocity and the output power data have frequently variable characteristics. Hence, the amount of wind energy to be generated will be closely related to the wind velocity in that region. The energy which can be generated on a wind turbine depends on several parameters such as the wind velocity, the tower height of the wind turbine, the wing diameter, the density of the zone, and the wind blowing duration.

Organic farming activities gained a worldwide momentum due to the fact that consumers in advanced countries have come to realize the damages caused by industrial processes to human health as well as environment. This type of production activity has spread rapidly all over the world and it has been observed that it was possible to yield a richer aroma in the products grown under organic farming conditions. Organic livestock activities have survived from the 1980s onwards. The main goal in this kind of activity is to restore the natural balance which tended to be lost in the ecosystem and to give weight to animal health, as well as human health [1]. Besides, industrial agriculture and stock raising generally require high energy consumption. This requirement brings about economic and environmental problems. In order to prevent these problems, the issue of agricultural and stock raising applications integrated with renewable energy have become an important research topic in recent years, particularly in European countries.

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Likewise its utilization in many other areas, artificial neural networks (ANN) have been used also for the determination of wind data such as wind velocity data and blowing duration of wind. By using statistical methods plus ANN, Kırbaş predicted the short-term multi-step wind velocity [2]. Using the wind velocity data for Akhisar region, Ata predicted the wind-blowing duration of that region with ANN [3]. Gnatowska and Moryn-Kucharczyk evaluated the wind energy perspective in Poland particularly in terms of wind power. He emphasizes the significance of state support in the production of wind energy in his study [4]. In an attempt to increase the estimation quality in wind energy studies, Yan and Ouyang suggested a new approach and verified their study in a real-world wind production facility [5]. In their study, which well summarized the characteristics of wind energy, Güngör and Eskin examined the deficiencies in previous studies, while also pointing out the possible future development areas in energy envisionings. [6]. In his study, Güngör investigated the effect of global warming on the potential of wind energy [7].

In this study, by using the wind velocity data at an altitude of 10 meters in Ardahan and Erzurum which was obtained from Turkish General Directorate of Meterological Affairs (GDMA) [8], wind blowing durations were calculated. Wind blowing durations obtained were predicted with ANN. The present study aims at integrating the renewable energy sources with agriculture and stock raising, and is intended to be served as an example for respective engineers, designers and planners in this regard.

### 2. AGRICULTURAL AND LIVESTOCK CHARACTERISTICS IN THE STUDIED REGION

The urbanization rate in Erzurum is below Turkey's average and rural employment in the city constitutes 62.3% of total employment. These socio-economic indicators suggest that stock raising is important for the provincial economy and will preserve this importance in the near future. Number of milked cattle in Erzurum constitutes 4.5% of the total milked cattle in Turkey, while its contribution to Turkey's total milk production amounts to 4.3% of the overall milk production in the country. The presence of sheep and goat in the province has demonstared a significant decline in recent years. As for 2017 data, there are 730.000 cattle and 714.000 sheep and goat in Erzurum [9,10].

Due to elevation features and air masses effective throughout the year, a harsh continental climate is dominating in Ardahan. Based on these parameters, the average temperature is below -10 °C in winter and is about 16-17 °C in summer period. According to the official data of the GDMA, the average annual temperature is 3.7 °C. The highest temperature in Ardahan is measured as 16.3 °C in July, whereas lowest temperature values are measured in January. In addition, due to its geographical location, the large mountain ranges in southwest-northeast direction are found to have an altitude about 2000-2200 meters whereas in volcanic masses it even reaches to 3000 meters and over. Due to the fact that the climatic conditions are convenient for agriculture, the proportion of land allocated to agriculture in the province is around 17%, 80% of which is assigned to grain products. In the remaining areas, the mostly cultivated product group is forage crops. Through some narrow areas in the vicinity of Posof and in Kurtkale basin which enable pomiculture, a variety of fruits are grown such as apples and walnuts [11].

Cattle raising is of particular importance in Ardahan and is mainly represented by beef raising. While there is more cattle in the central district and Göle with their larger grassland areas, in the villages of Posof basin the number is much fewer due to lack of pasture areas. Small livestock in the province is basically represented by sheep raising. It is mostly raised in the central district and Çıldır. Poultry farming, which is mainly forefronted by goose breeding, is a city-wide common activity. Other poultry includes chicken, turkey and duck [11].

Considering the agricultural and livestock activities and the potential in the region, it can be suggested that these activities can be increased through several governmental promotions and support. Rough and high mountainous areas in the region is one of the main factors restricting the effective use of region's available lands. Besides, agricultural and stock raising activities when supported by future wind power plants in the region shall contribute to the efficient use of the area.

## 3. DESIGN

There are varios methods available in the literature to determine the wind energy potential of a region, yet, the most commonly used ones are the Weibull and the Rayleigh distributions. In order to determine the duration and distribution of the wind in a region during a given period, in addition to the Weibull distribution, Rayleigh distribution is used by taking the shape parameter in the probability density function as two. Generally, the Rayleigh distribution is used when only the average velocity value is known for the region.

Rayleigh density function is shown as below;

$$f_R(v) = \frac{\pi}{2} * \frac{v}{v_{ort}^2} \exp\left(-\left(\frac{\pi}{4}\right) * \left(\frac{v}{v_{ort}}\right)^2\right)$$

(1)

Rayleigh cumulative distribution function is shown as below;

$$F_R(v) = 1 - \exp(-\frac{\pi}{4} (\frac{v}{c})^2)$$
<sup>(2)</sup>

where; c is parameter of scale (m/s). The biggest advantage of the Rayleigh distribution function is that it can determine the wind distribution of a given region depending only on the average velocity. The blowing velocity/hour can be calculated according to the Rayleigh distribution function with the help of Equation 3 [12].

$$h_r = f_R(v) * 8760$$

(3)

The average annual wind velocity data for Ardahan and Erzurum at altitude of 10 m for the years 2014-2018 are given in Table 1 and 2, respectively [8]. Out of five-year histrorical data regarding Ardahan, the highest average wind velocity is featured in station no.3 (Çıldır) with 3.515 m/sec. The lowest average wind velocity featured in the same city belonged to station no.2 (Posof) with 0.951667 m/s. In Erzurum, while the station no.25 (Karaçoban) is observed as having the lowest wind velocity with 1.41 m/s, the highest value appeared in station no.12 (Yakutiye) with 4.180 m/s.

Table 1. The average annual wind velocity data for Ardahan [m/s] [8]							
Measurement	Province/region	2014	2015	2016	2017	2018	
station							
1	Ardahan	1.25	1.283333	1.025	1	0.991667	
2	Posof	0.866667	0.883333	1.1	0.991667	0.916667	
3	Çıldır	3.683333	3.383333	3.65	3.425	3.433333	
4	Damal	2.466667	2.575	2.7	2.608333	2.516667	
5	Göle	2.108333	2.141667	2.216667	2.083333	2.091667	
6	Hanak	2.025	2.091667	2.116667	2.208333	2.05	

 Table 2. The average annual wind velocity data for Erzurum [m/s]
 [8]

	Province/region	2014	2015	2016	2017	2018
station 7	Erzurum	1.458333	1.391667	1.466667	1.341667	1.391667
8	Erzurum Airport	3.083333	2.783333	3.4	3.025	2.575
9	Uzundere	1.666667	1.791667	1.783333	1.8	1.883333
10	İspir	1.666667	1.716667	1.741667	1.641667	1.783333
11	Oltu	2.175	2.133333	2.058333	2.125	1.975
12	Yakutiye	4.857143	4.75	4.714286	3.166667	3.414286
13	Palandöken	4.033333	3.4125	3.366667	3.641667	3.77
14	Aziziye	4.011111	9.74	3.022222	2.755556	2.366667
15	Tortum	1.441667	1.491667	1.483333	1.541667	1.425
16	Horasan	1.141667	1.2	1.15	1.166667	1.233333
17	Aşkale (Kop Mountain)	3.375	3.7	4.036364	3.2	3.025
18	Hinis	1.258333	1.35	1.4	1.308333	1.416667
19	Palandöken Ski Center	4.866667	3.4	3.95	3.85	4.033333
20	Pasinler	4.008333	3.916667	2.833333	2.375	2.266667
21	Narman	1.908333	1.85	1.925	1.958333	1.808333
22	Çat	3.358333	3.916667	2.683333	2.308333	2.291667
23	Karayazı	3.125	3.25	3.391667	3.175	3.175
24	Aşkale	1.35	1.525	2.275	2.183333	2.083333
25	Karaçoban	1.233333	1.183333	1.083333	1.025	1.358333
26	Köprüköy	2.375	2.291667	2.45	2.2	2.216667
27	Olur	2.008333	2	2.066667	2.1	2
28	Pazaryolu	2.641667	2.55	2.6	2.35	2.583333
29	Şenkaya	2.525	2.55	2.558333	2.633333	2.5
30	Tekman	2.633333	2.391667	2.641667	2.35	2.375
31	Çayırözü	2.428571	2.533333	2.166667	2.2	2.391667
		=				

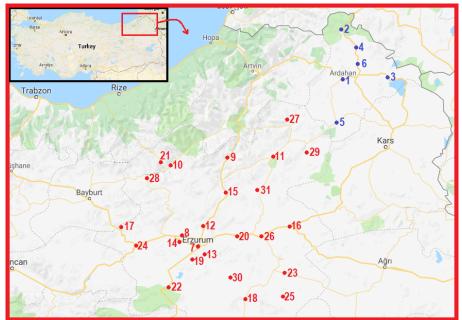
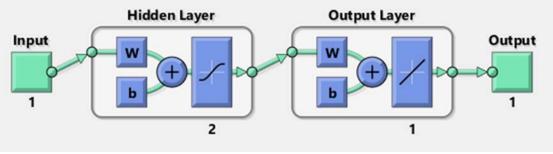


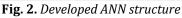
Fig. 1. Locations of the measurement stations

Fig.1 shows the locations of the measurement stations from which the data is collected. Blue and red colors demonstrate the stations in Ardahan and Erzurum, respectively. As can be followed clearly from the figure, the measurement stations cover a satisfactorily large area and can well generate comprehensive results about the regions under study.

## 4. SOLUTION METHOD

In this study, feedback propagation learning algorithm was used to determine the wind blowing duration. The study benefited from the artificial neural network toolbox which is available in MATLAB/Simulink software package. Normalization techniques were applied to input values in order to improve the performance results. The input and output data of the ANN which was obtained from the GDMA consist of 70% training cluster, 15% validation cluster and 15% test cluster, which amounted as 1249, 268 and 268 data figures, respectively. The performance analysis of the neural network was evaluated by the Mean Square Error (MSE) and the Regression curve. According to the test results, the best results for wind blowing duration prediction were obtained via 9 hidden layer architecture. Network was trained using the Levenberg-Marquardt optimization technique. Using input/output values, different network configurations are trained in order to produce minimum Mean Square Error (MSE) and maximum regression values. The ANN architecture designed for the study is given in Fig. 2.





## 5. RESULTS AND DISCUSSION

The results indicate that, the station with the maximum wind duration in Ardahan is Posof, while the station with the least blowing duration is station no.3. Similarly, the maximum and minimum wind blowing duration in Erzurum belong to station no.25 and no.12, respectively. Maximum wind blowing durations occured in the areas which had minimum wind velocity; and, just in contrast, minimum wind blowing durations occured in the areas with maximum wind velocity. Table 3 and Table 4 depict the annual average values (in hours) of wind blowing durations produced via Equation 3. It can be observed that wind-blowing durations for the areas under study vary between 479.6-5825.7 hours and 1643.6-6710.8 hours per year for Erzurum and Ardahan, respectively.

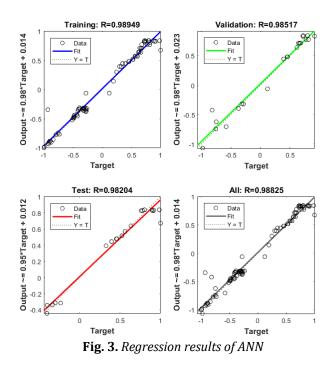
Measurement station	Province/region	2014	2015	2016	2017	2018			
1	Ardahan	4733.203	4370.249	5756.821	5658.453	6041.088			
2	Posof	6495.56	6710.821	5324.31	5846.948	6345.464			
3	Çıldır	1646.835	1805.675	1643.636	1765.14	1755.765			
4	Damal	2337.085	2343.559	2213.281	2297.478	2413.2			
5	Göle	2763.522	2747.942	2693.202	2814.068	2909.252			
6	Hanak	2840.275	2869.319	2810.844	2667.931	2917.438			
Table 4. Annual average values (in hours) of wind blowing durations of Erzurum (hrs/year)									
Measuremen			2015	2016	2017	2018			
station	, 0								
7	Erzurum	4145.23	8056 4165.	447 4000.046	4268.346	4250.11			
8	Erzurum Airpo	ort 1955.67	7937 2087.	265 1765.158	8 1871.672	1260.116			
9	Uzundere	3143.11	204 2989.	538 3070.985	2997.615	2973.711			
10	İspir	3582.86	5985 3414.	167 3440.17	3636.309	3379.16			
11	Oltu	2811.79	328 2887.	914 2995.521	2871.655	3129.309			
12	Yakutiye	743.355			1387.385	1200.451			
13	Palandöken	1450.10	0758 1214.	802 1841.948	1699.523	1338.103			
14	Aziziye	1034.91	279 120.0	423 1604.926	2285.331	2373.665			
15	Tortum	4244.62	.423 3961.	477 4089.219	4035.151	4257.486			
16	Horasan	5114.41	945 4791.	188 5119.905	4967.086	4946.985			
17	Aşkale	1628.93	8456 1393.	711 1407.308	8 1450.408	1999.496			
18	Hınıs	4671.12		484 4294.659	4632.293	4299.778			
19	Palandöken	1225.52	2058 154.1	213 1562.849	1615.985	1523.813			
20	Pasinler	1497.82	276 1530.	157 2064.185	2508.328	2658.609			
21	Narman	3066.36	5423 3144.	852 3124.87	3021.858	3351.015			
22	Çat	1807.19	975 1507.	764 2172.554	2555.485	2652.498			
23	Karayazı	1941.27	7699 1906.	42 1804.562	1918.092	1918.092			
24	Aşkale	4340.22	2092 3854.	256 2625.005	2747.386	2900.057			
25	Karaçoban	4716.85	i 948 4982.	754 5567.42	5825.745	4377.563			
26	Köprüköy	2458.29	9597 2518.	716 2428.775	2592.623	2717.93			
27	Olur	2885.87	709 3016.	304 2879.499	2834.338	2985.84			
28	Pazaryolu	2191.18	386 2251.	299 2221.07	2411.023	2326.168			
29	Contraria	2313.40	)7 2319.	602 2362.507	2279.923	2444.502			
	Şenkaya	=010110							
30 31	Tekman	2228.62			8 2441.056	2539.75			

**Table 3.** Annual average values (in hours) of wind blowing durations of Ardahan (hrs/year)

**Table 5.** Regression and MSE for different architecture.

Hidden	Regression		<i></i>		Encoh
Layer	Training	Validation	Test	MSE	Epoch
2	0.9894	0.9851	0.9820	0.011	800
3	0.8918	0.9315	0.8800	0.017	800
4	0.8874	0.9232	0.9366	0.016	800
5	0.9001	0.8879	0.9239	0.024	800
6	0.8943	0.9247	0.9097	0.018	800
7	0.8951	0.8869	0.9360	0.026	800
8	0.9001	0.8970	0.9156	0.025	800
9	0.9048	0.8721	0.9112	0.026	800
10	0.8951	0.8932	0.9042	0.028	800

A given regression result converging to 1 represents a close relationship between variables, while they are meant to have a weaker relationship when the regression result converges to 0. As shown in Fig. 3, the results of the network with 2 hidden layer have a regression value of approximately 1, which means that ANN successfully predicts the unknown data.



The output function obtained as a result of ANN modeling and prediction of the data is provided in Fig. 4, and error histogram and validation MSE obtained from the study are given in Fig. 5. As expected, maximum error values shown in Fig. 5 are noted to occur during the training of the network.

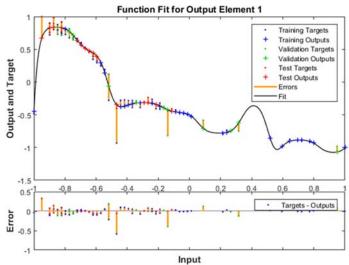


Fig. 4. Simulation and Real Values of Data

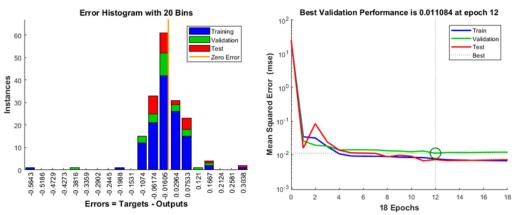


Fig. 5. Error histogram of ANN modeling (left), Validation MSE values (right)

#### 6. CONCLUSIONS

As a renewable energy source, durableness of wind energy and pre-measurements for turbine installation are of great importance. The installation of the wind turbine at an optimal location is a parameter that directly affects the amount of electrical energy to be obtained. In this study, wind blowing durations in Erzurum and Ardahan, two eastern cities in Turkey, where cattle raising and agriculture activities are carried out, were determined with Rayleigh distribution and the results were predicted by artificial neural networks which is one of the techniques of machine learning. Prediction results performed a 99% accuracy. The current study showed that agriculture and stock raising activities would be carried out in integration with wind energy, especially for the locations with long wind blowing durations. Considering the high performance rates as for the prediction of artificial neural networks, this method is promising in predicting several other parameters regarding wind energy.

As far as the regions under study is concerned, it is possible to assert that these regions can be improved in terms of energy as well as agriculture and stock raising. Integrating agriculture and stock raising activities with wind energy, which is a sort of renewable energy, would yield to increase the efficiency and effectiveness and make favorable contributions to environment.

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