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# Evaluation of wind power potential in shelek corridor (Kazakhstan) using weibull distribution function

**Abstract.** Kazakhstan currently has one of the highest, per capita, carbon footprints in the world. There are heavy reliance (approx 85%) on coal for electricity production in Kazakhstan. Coal is a very carbon intensive fuel. A drive to moderate coal's contribution to electricity production provides a driver for wind energy development. Finding a suitable location requires a detailed and often costly analysis of local wind conditions. Wind resource assessment is a crucial first step in gauging the potential of a site to produce energy from wind turbines. In this paper, the wind energy potential of Shelek corridor, located in the Almaty region in Kazakhstan were examined. Local wind speed distributions are represented by Weibull statistics. The results show that the average annual mean wind speed variation for Shelek corridor ranges from 4.0 to 8.0 m/s. The wind power density variation based on the Weibull analysis ranges from 280.0 to 320.0 W/m<sup>2</sup>.

Key words: wind power, energy production, renewable energy, MERRA, Weibull distribution.

## Introduction

For nowadays Kazakhstan is among the top 20 countries which produce the highest carbon dioxide emission per capita. There are several reasons of this such as a small population per square kilometer, a strong dependence of the power plant on coal. Coal emits very high amount of carbon dioxide to the atmosphere. It means that wind electric stations will be attractive opportunity for Government and business sector. Investing in such projects is commercially feasible [1].

Kazakhstan's geographical location is very suitable to develop wind energy stations with power up to 760 GW. For instance; Shelek corridor has potential to produce power from wind energy aprroximately 3200 kW/h/MW [2].

The performance of the wind power conversion system depends on a wide range of criteria including

a lot of aspects [3]. The characteristics of the wind tendency are one of the main and influential parameters that support when determining the suitability of a site for the use of wind energy. Due to the stochastic nature of the wind, several models were developed to analyze the available wind data in order to obtain the characteristics of any wind mode. One of these established models operates on the basis of the Weibull distribution [3]. This article presents the details of the analysis based on the Weibull distribution of wind data, from the NASA's Modern-Era Retrospective Analysis for Research and Applications.

#### Kazakhstan's wind power market

The Republic of Kazakhstan has the ninth largest territory in the world. It is about 2.7 million km<sup>2</sup>. However, Kazakhstan is one of the countries

with the least population density in the world  $(5.5 \text{ persons per } \text{km}^2)$ . There are three cities with population over 1 million people: Astana, Almaty and Shymkent.

Kazakhstan has huge resources of fossil fuels such as coal, oil, gas and uranium. These resources are being actively exporting. For instance, Kazakhstan produces just less than 100 million tons of coal and approximately 35% of this is exported to the neighboring countries. 1.5 million barrels of oil produced per one day. And 75% of them sold to the China and Russia.

There is no any country in the world, which produces more uranium than Kazakhstan. According to the data in 2009 year, 27.6% of world's uranium production was made by Kazakhstan.

Excess of energy resources is the main factor hindering the development of renewable energy sources in Kazakhstan.

Based on the outcomes of the World Summit on Sustainable Development held at Johannesburg in 2002 Kazakh government accepted Sustainable Development Concept for 2007-2024. That Concept was devoted to sustainable use of renewable resources and alternative energy in Kazakhstan. Efficient and sustainable use of renewable resources and alternative energy sources will be possible if:

 innovative technologies in use of land, water, forestry, fishery, biological resources and renewable sources of energy will be introduced;

 efficient use of hydropower resources, solar and wind resources and other renewable resources and alternative energy sources will be simulated;

- the Centers for distribution of international experience in the area of energy and resource efficiency and use of renewable sources of power will be established.

Considering all of the above, it is possible to list the main challenges, which are the strong drivers for the development of wind energy in Kazakhstan:

- the old power generation infrastructure;

- high amount of transition and distribution loses;

- The highest carbon footprints in the world.

As a result, since 2012 year the amount of installed capacity by the onshore wind turbines has increased 65 times. It can be shown in the diagram below.



Figure 1 – Trends in renewable energy (Installed capacity)

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In 2012, installed capacity was only 1.5 MW, the first wind power station was installed in Kordai, Zhambyl region. Next year power production increased three times. 2014 year was significant for Kazakhstan wind power market, because the second wind power station was set up in Ereimentau, Akmola region. And it drove to fundamental growth in the installed capacity: from 5.5 MW in 2013 to the 52.81 MW in 2014. In 2015 and 2016, there were still positive trends in the installed capacity. Compared to 2014 year, potential power produced from the wind rose by the 40% each year.

#### Methods and materials

There are two ways of obtaining data. First way is analysing of production data. The benefits of using production data are that they are reflecting true fluctuations and do not require any additional calculations. But this approach is applicable to the places where a wind turbine has already been installed. The second way is using data from the weather station. However, that way has several disadvantages: it is not available; data records sometimes not complete. Moreover to that, to calculate a new location, you need to install anemometers. This is not economically profitable [4].

In Kazakhstan there is no free available wind speed data at the moment. Companies should measure wind data themselves. Previously a wind atlas of Kazakhstan was created in the framework of the UNDP / GEF Project and the Government of the Republic of Kazakhstan "Kazakhstan is an initiative to develop the wind energy market". However, at this moment that site is not available.

Ritter et al. [4], recommended an alternative dataset which is providing wind power analysis and reanalysis data, such as Modern-Era Retrospective Analysis for Research and Applications (MERRA) data provided by NASA [5]. MERRA offers wind data all over the world and an hourly temporal resolution since 1979. It consist two components at three different heights (2 m, 10 m and 50 m above ground). A northward and an eastward wind component are helpful to derive the wind speed and wind direction at various turbine heights [6].

In this study the data used from "MERRA-2 inst1\_2d\_asm\_Nx: hourly, instantaneous, Single-Level, Assimilations, Single-Level Diagnostics V5.12.4" during the period from 01.01.2015, until 28.02.2018 for each day. We used the northward and eastward wind speed at the heights of 2 m, 10 m and 50 m above the ground. To cover all Shelek corridor grid points with a latitude between 440 E and 47.50 E and a longitude between 73.50 N and 77.80 N are used.

In recent years, more attention has been paid to the Weibull distribution, as suggested by the nearby approximation of the probabilistic laws of a number of natural phenomena and is expected to ensure a good correspondence to the experimental data [7].

Mathematically, i.e. The Weibull distribution function of the two parameter functions is expressed as:

$$f(v) = \frac{k}{c} (\frac{v}{c})^{k-1} e^{-(\frac{v}{c})^k}$$
(1)

The integral distribution function F(x) is given by:

$$F(v) = 1 - e^{-(\frac{v}{c})^{k}},$$
 (2)

where:

v – wind speed, m/s;

k – shape factor ;

c – scale parameter, m/s.

The shape factor k is the main factor in determining the uniformity of the wind. The uniformity of the wind changes direction with a change to i. with increasing k, the uniformity of the wind increases.

It has been established that almost all parameters, such as wind speed, wind speed probability, must be within a certain range, the energy is available in a certain mode, etc., are necessary to fully appreciate the dignity of the wind regime, depend on the rough calculation of these values of k and c [8].

k and c parameters can be found with the several methods such as the power density method; least square method, Modified likelihood method etc. [9]. The methods are briefly discussed below

## 1) Standard deviation method

The standard deviation method gives the ratio between the mean  $(v_m)$  and the standard deviation  $(\sigma_v)$ :

$$\left(\frac{\sigma_{\nu}}{\nu_{m}}\right)^{2} = \frac{\Gamma\left(1+\frac{2}{k}\right)}{\Gamma^{2}\left(1+\frac{1}{k}\right)} - 1$$
(3)

After *k* is determined, c is determined as:

$$c = \frac{v_m}{\Gamma\left(1 + \frac{1}{k}\right)} \tag{4}$$

In a simple methodology, an acceptable approximation for k and c can be approximated:

$$k = \left(\frac{\sigma_v}{v_m}\right)^{-1.090} \tag{5}$$

$$c = \frac{2v_m}{\sqrt{\pi}} \tag{6}$$

## 2) Empirical method (EM)

A special case of the method of moments, the empirical method determines k & c, as in equations (4), (5):

$$k = \left(\frac{\sigma}{\bar{v}}\right)^{-1.086}$$
$$c = \frac{\bar{v}}{\Gamma\left(1 + \frac{1}{k}\right)}$$
$$\Gamma(x) = \int_0^\infty t^{x-1} e^{-t} dt$$

#### 3) Power density method (PDM)

This is the one of the most important methods of determining k and c. First of all, energy pattern factor  $E_{pf}$  is calculated by the division of cubic wind speed to cube of mean wind speed [9-10]. After that k and c may be found by formulas (7, 8):

 $k = 1 + \frac{3.69}{E_{pf}^2} \tag{7}$ 

$$c = \frac{\bar{\nu}}{\Gamma(1+\frac{1}{\bar{\nu}})} \tag{8}$$

#### 4) Modified likelihood method (MLM)

That method was proposed by Stevens and Smulders [10] and describes k and c by the following formulas:

$$k = \left[\frac{\sum_{i=1}^{n} v_{i}^{k} ln(v_{i})}{\sum_{i=1}^{n} v_{i}^{k}} - \frac{\sum_{i=1}^{n} ln(v_{i})}{n}\right]^{-1}$$
$$c = \left(\frac{1}{n} \sum_{i=1}^{n} v_{i}^{k}\right)^{1/k}$$

## **Results and Discussion**

In this study the data used from 01.01.2015, until 28.02.2018 for each day. The northward and eastward wind speed at the heights of 2 m, 10 m and 50 m above the ground were horizontally interpolated by the formula [4]:

$$v_m = \sqrt{u_h^2 + v_h^2} \tag{9}$$

Monthly mean values of wind speed are shown in the tables 1-3. The frequency distribution of the data is shown in Table 4.

Table 1 – Monthly mean velocities at height of 2 m

Month	Mean velocity m/s					
	2015	2016	2017	2018		
January	2.7242	3.3995	2.9572	5.9183		
February	3.4051	3.6068	2.8878	3.0978		
March	2.2784	2.8207	3.4395	-		
April	2.0906	2.2409	1.6123	-		
May	1.5377	2.8716	2.6959	-		
June	2.4187	1.0863	0.8440	-		
July	2.4385	1.3534	2.4462	-		
August	1.9352	5.6995	4.1758	-		
September	2.5953	5.6995	2.6755	-		
October	2.0809	2.5965	2.1037	-		
November	2.2991	2.3995	2.5797	-		
December	2.9105	2.5473	3.1505	-		

Month	Mean velocity m/s					
	2015	2016	2017	2018		
January	3.3320	3.6725	3.3897	6.4588		
February	3.8281	4.0027	3.4253	3.6344		
March	2.9956	3.2741	4.0367	-		
April	2.3681	2.6763	2.0381	-		
May	1.8185	3.0655	2.9150	-		
June	2.8588	1.2251	0.9799	-		
July	3.0926	1.4276	2.9774	-		
August	2.4606	6.4904	4.7401	-		
September	3.0329	6.4904	3.3571	-		
October	2.4876	3.0643	2.6479	-		
November	2.7427	3.0442	2.9889	-		
December	3.4514	2.9444	3.7378	-		

Table 2 – Monthly mean velocities at height of 10 m

Table 3 – Monthly mean velocities at height of 50 m

Month	Mean velocity m/s				
	2015	2016	2017	2018	
January	5.4955	5.8329	5.8604	9.8087	
February	5.8815	6.2380	5.3924	5.6900	
March	4.3439	6.3175	6.0201	-	
April	3.7365	4.8102	3.1457	-	
May	2.4771	4.1304	4.5861	-	
June	4.7340	4.1634	1.7987	-	
July	4.9166	1.6786	4.7932	-	
August	4.0327	2.2848	7.8409	-	
September	4.7945	9.3863	5.1376	-	
October	3.8964	9.3863	4.1561	-	
November	4.6451	5.0147	4.9872	-	
December	5.4955	4.5663	6.0211	-	

According to these table the highest speed was detected in January, 2018 and reached almost 6m/s in between 2015 and 2017, and over 9 m/s in 2018. Respectively in the beginning of summer there always was low wind speed. The least meaning of wind speed for summer is 0.9 m/s.

Collected speed data was also divided to the intervals to find more common velocity. It was observed that the most frequent velocity at height 2 m is  $2\div3$  m/s (0.58). As for 10 m height  $3\div4$  m/s repeated with almost 40% frequency. The wind

speed of 4-6 m/s at the height of 50 m was repeated at a frequency of 0.6.

As we can see from the table, when the height is increasing the value of velocity is also increases. Thus, wind speed data at these three heights could be vertically extrapolated to the turbine height [11].

From the above formulas and wind data, the results for shape and scale parameters are calculated using various methods described above is indicated in Table 4.

	Frequency					
Methods	2 m		10 m		50 m	
	k	с	k	с	k	с
SDM	2.74	3.14	2.93	3.66	3.10	5.74
EM	2.53	3.13	2.70	3.65	2.86	5.72
PDM	2.73	3.12	2.92	3.63	3.09	5.70
MLM	2.62	3.13	2.76	3.63	2.93	5.70

Table 5 – Wind velocity distribution

Accuracy of calculations was checked by the following three methods: RMSE, R2 and Chi - Square tests:

$$RSME = \left[\frac{1}{N}\sum_{i=1}^{N}(y_i - x_i)^2\right]^{1/2}$$
$$\chi^2 = \frac{\sum_{i=1}^{N}(y_i - x_i)^2}{N - n}$$

$$R^{2} = \frac{\sum_{i=1}^{N} (y_{i} - z_{i})^{2} - \sum_{i=1}^{N} (y_{i} - x_{i})^{2}}{\sum_{i=1}^{N} (y_{i} - z_{i})^{2}}$$

Where  $y_i$  is observed frequency and  $x_i$  Weibull's frequency, N is a number of observations, n is a number of used constants. The results are shown in a table below

Table 6 – Wind velocity distribution

Tests	SDM	EM	PDM	MLM
RSME	0.675	0.0678	0.0692	0.0673
Chi-square	0.063	0.0063	0.0067	0.0061
R <sup>2</sup>	09485	0.9469	0.9472	0.9466

As shown in Figures 1-6, all the methods discussed show more or less similar results, even if the maximal likelihood method and the energy properties are a little more accurate and, therefore, can be considered as the most appropriate.

As we can see from the graph, the most probable wind speed at this height is under 3 m/s. And our calculations have shown the mean value of the velocity is 2.78 m/s.

According to the Fig.4 we can see that the most probable wind speed occurs at speed 5 m/s with a probability of 22.5%. This means that at Shelek corridor the wind speed that often arises at 5 m/s.

Usually, 4-5 m/s is an ideal wind speed for wind turbines. However, the use of wind energy is commercially installed only for high (8-9 m/s) and medium (6-7 m/s) wind conditions. If wind energy

is available in low wind conditions, it is possible to develop turbines specifically for these regions, which will help to reduce dependence on fossil fuels.

To analyze the wind energy potential there is also mean power density based on Weibull distribution must be calculated [13].

$$P_{WD} = \frac{1}{2}\rho c^{3}\Gamma(1+\frac{3}{k})$$
(10)

Where *c* shape factors which were calculated by the four methods described below.  $\rho$  is air density related to the pressure, temperature and humidity. However, air density has not significant effect on wind resource calculations, thus it can be taken as a constant value 1.225 g/cm<sup>3</sup>.



Figure 2 – Weibull distribution at the height of 2 m



Figure 3 – Weibull distribution at the height of 10 m





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On the other hand, mean power density can be calculated by the formula:

$$P_{REF} = \frac{1}{2}\rho\bar{v}^3 \tag{11}$$

# Conclusion

The potential of wind energy in Shelek corridor have been studied in this analysis using widely used Weibull distribution technique. The Weibull parameters such as shape factor and scale factor have been calculated by four different methods such as least squares method, power density method, empirical method and modified likelihood method. Relative percentage of error and chi-square error has been analyzed for each method and also calculated the efficiency of these methods. The wind data was obtained from NASA Modern-Era Retrospective Analysis for Research and Applications "MERRA-2 inst1\_2d\_asm\_Nx: hourly, instantaneous, Single-Level, Assimilations, Single-Level Diagnostics V5.12.4" during the period from 01.01.2015, until 28.02.2018 for each day.

The results found that Shelek complex is prospective site to set up vertical axed wind turbine. The results have been verified by Weibull distribution technique where Weibull shape factor and scale factor were calculated using four different approaches. The statistical analysis also found the modified likelihood method is more efficient method with minimum error in the wind data analysis. The study presented the potential of Shelek corridor to produce pure energy using wind power.

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