

LSTM Model Based Wind Speed Forecasting

S. Chandrababha¹, C.Dinesh², A.Mohamed Ibrahim³ & G.Saravanan⁴

^{1,2,3,4}Department of Electrical and Electronics Engineering,

KPR Institute of Engineering and Technology, Coimbatore, Tamil Nadu, India.

Email: chandrababhasbe@gmail.com¹

Abstract

The forecasting of wind speed is significant factor. The wind energy is the hygienic form of fuel that has an effect on global warming and impact on worldwide climate. The perceptions are met on the ground truth that the energy condition of wind speed which is infers an immediate connection to the wind speed. Reliable forecasting of wind speed is much more crucial in the preparation of the electricity grid and the planning of a power system. Especially regression analysis of linear nature and LSTM procedure followed to anticipate one year power generation from wind energy. The model built with the three years real time data collected from Tamilnadu (Coimbatore).The proposed model encouraging in the perception of ground truth and low contrasted with direct relapse model which is around 2.0448.

Keywords: Deep Learning, LSTM, Machine Learning in Wind speed forecast, Regression Algorithm.

1. Introduction

Inexhaustible sources like breeze, sun powered are quickly utilized. The vision of nation to commission 175 GW power from natural sources such as hydro (5 GW), biomass (10 GW), wind (60 GW) and solar (100 GW) before the year 2022. The association of renewable energy established 37.09 GW of wind power generation before the month of October 2019. The 70 % electricity generation relies on fossil fuels and releases CO₂, SO₂, NO_x about 2.5 BT, 20 MT and 8 MT respectively during translation process which is destructive to the environment. It has had an adverse effect on the environment and the adoption of green energy does seem to have a global impact. Fuel sources like sun based and wind are relied upon to give added implication in arranging of energy prerequisites everywhere on the nations and particularly wind energy plays further realistic. Breeze power consumption is expected to grow in the global system and power valuation is a challenging attribute. Thusly exceptionally exact breeze speed expectation is undoubtedly necessitated that helps in having diminished unfriendly impact of wind ranch. It is exceptionally huge for the improvement of wind power [5].

There are numerous customary strategies like physical techniques, measurable techniques, ordinary techniques, neural organization, mixture strategies which consolidates at least one strategies so as to improve the precision of the breeze speed expectation calculation. The AI, DL and ML processes are popularly applied for wind speed forecasting [2]. The LSTM model is proposed here and now AI calculation built which all together for exact breeze speed expectation in grid. Prediction could be categorized into four ways based on the time horizons: Sec to Hr, Hr to Hr, Hr to days, days to Year [9]. Precise guess of wind speed will progress power perception rate and create noble impression.

2. Forecast Dataset

The performance of breeze power is calculated by $P=0.5\rho Av^3$. Here P is output power, ρ is air density. v is the velocity of wind, A is wind mill surface normal to the wind. The 3-major factor (i) blade radius (ii) velocity (iii) air density exists in energy generation and therefore more electricity is extracted at faster rpm and vice versa. The electricity obtained from kinetic energy principle from wind mill and the real time data (4 Years) collected from private wind farm organization[8][13]. This data set plotted in figure 1.

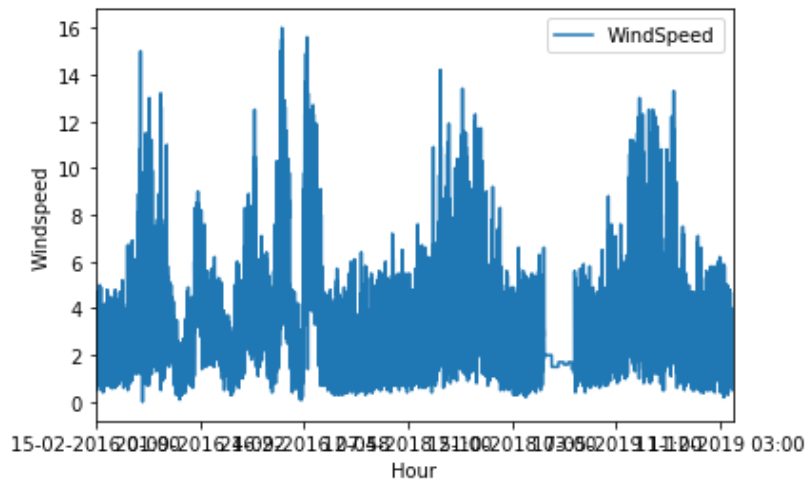


Figure 1. Wind Speed Hourly Data

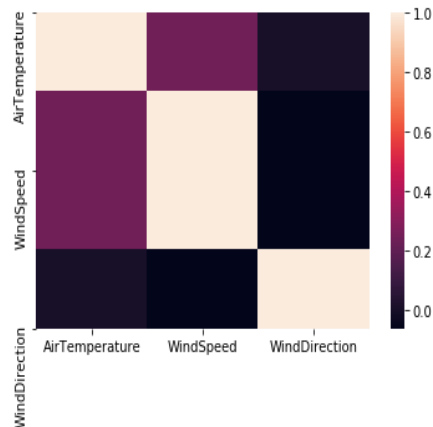


Figure 2. Heat Map

The 3-year information gathering (Hourly) between Feb-2016 and Dec-2018 used in this work and obtained at location (1.0259°, 76.9781°). The data have always been pre-processed and made up of 75% of the training and the remaining portion used for the test procedure. The model aspirations are assembled using regressive analysis (linear) and LSTM to estimate the upcoming 7650 hrs of breeze speed data [11]. The data set plotted as heat map as shown in figure 2 and figure 3 shows the wind speed with air temperature.

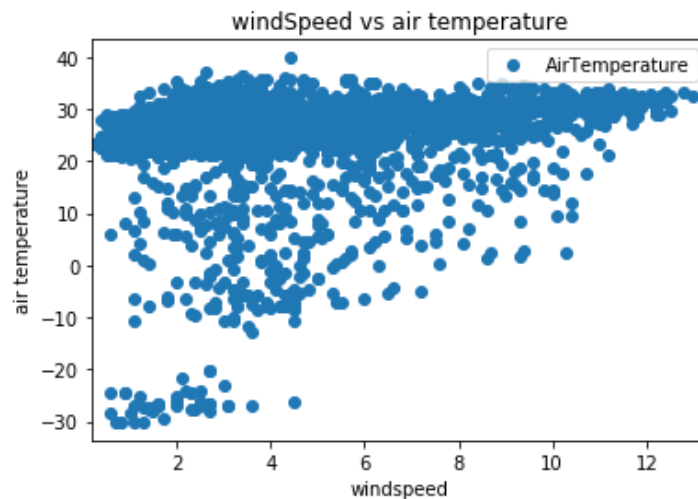


Figure 3. Wind Speed

3. Result and Discussion

System Information got as .xls is changed over into .csv (comma-isolated qualities) design for predisposing the verifiable information wherein the necessary information by methods for representation strategy, the library files are added, measurable feature such as size and shape is realized [10].

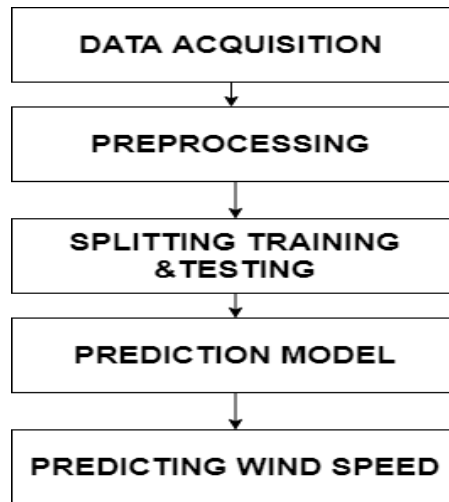


Figure 4. Flow chart for prediction of Wind speed

Following stage is to separate the informational collection into characteristics and names, where ascribes are autonomous factors and names are needy factors whose qualities are anticipated, the test size is 0.25 in the wake of parting informational index preparing must be finished by bringing in forecast model, fitting capacity that best fits the informational collection[7]. Testing is finished by methods for test informational collection, the calculation that precisely predicts up and coming 4-years. Stream chart of breeze velocity expectation appeared in figure 4. The regression (Linear) which is one of the managed AI calculation. The term relapse alludes to the way that we are attempting to anticipate such a proceeds with esteem number [14]. Direct relapse predicts a genuine worth yield dependent on the information. Slope plummet calculation is utilized [3]. The temperature and wind speed as variable and maps with each other. Direct relapse is agreed by the articulation $y(x)=m_0+m_1x$. here $y(x)$ is function of hypothesis and the plot is shown in figure 5. The foreseen and real consequences are displayed in figure 6.

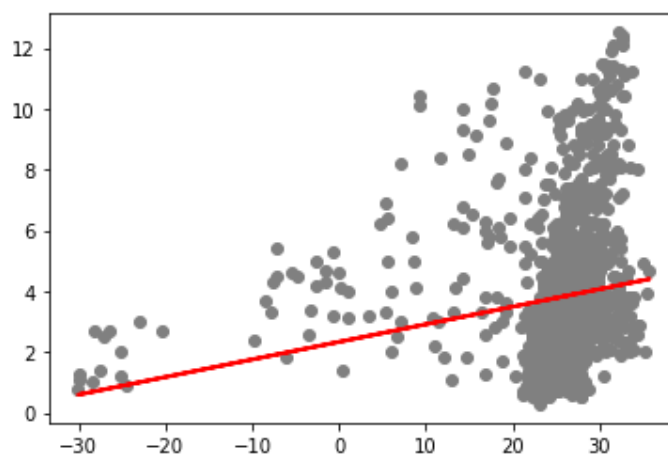


Figure 5. Regression Analysis (Linear)

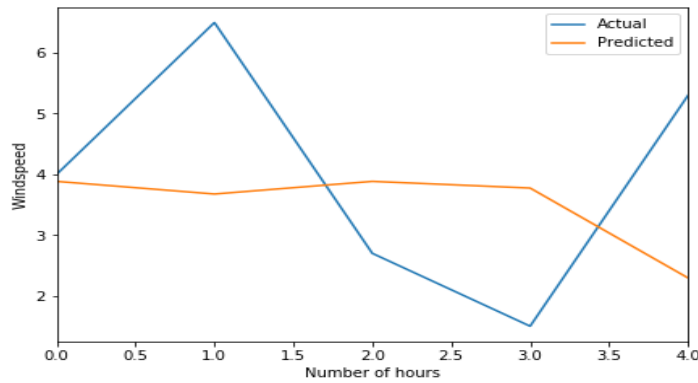


Figure 6. The Predicted Vs Actual Speed

The LSTM is the profound learning calculation which is an utilization of AI calculation. Intermittent neural organizations experienced momentary memory if a grouping is huge enough as it experiences the disappearing slope issue it later quits learning later LSTM was made as an answer for transient memory [12][6]. They hurl interior system called doors. The design of LSTM is appeared in figure 7.

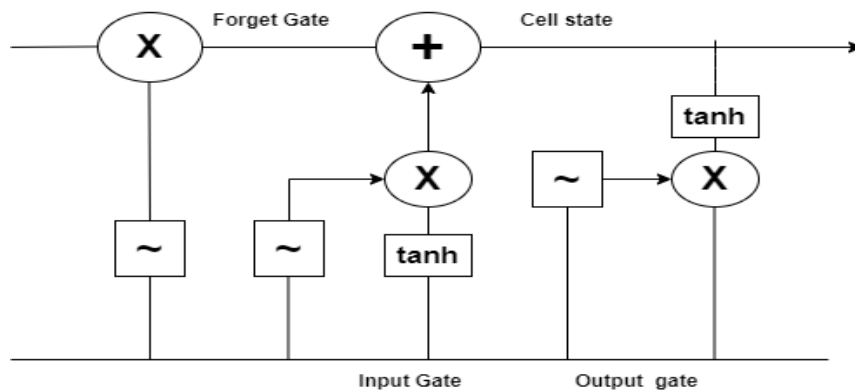


Figure 7. Architecture of LSTM

The gates can realize which information is significant for handling and which information isn't significant and it discards through overlook door. The cell express that goes about as a vehicle interstate that moves relative information.[1] This has 2 actuation capacities tanh and sigmoid enactment work. The tanh actuation stands utilized toward accomplish the informational collection via organization. The figure 8 illustrations of yield anticipated and outcomes of the developed framework are confirmed by the expressions RMSE, MAE and MSE [15].

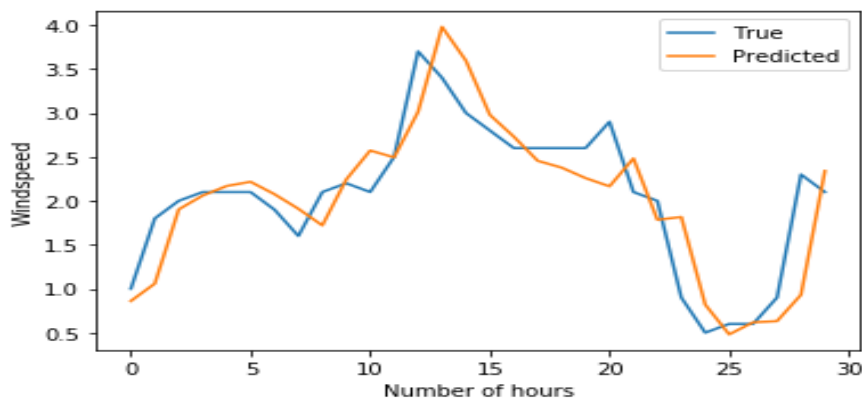


Figure 8. Predicted Vs True Speed

$$RMSE = \sqrt{\left[\frac{1}{N} \sum_{j=1}^N |y_j - \hat{y}_j|^2\right]}$$

$$MAE = \left[\frac{1}{N} \sum_{j=1}^N |y_j - \hat{y}_j|\right]$$

$$MSE = \left[\frac{1}{N} \sum_{j=1}^N |y_j - \hat{y}_j|^2\right]$$

Table 1. Model Validation

Methods	RMSE	MAE	MSE
Regression Analysis(Methods)	2.04	1.50	4.18
LSTM	0.90	0.65	0.82

4. Conclusion

Wind speed forecast model is the most noteworthy part in assessment of wind power framework that includes in arranging another wind mill. The expectation of wind speed model utilizing direct relapse calculation is contrasted and long momentary memory calculation utilizing the verifiable informational index of around three years and by approving utilizing execution measure [4]. It is discovered that the anticipated speed utilizing LSTM has precise than other model.

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