

| Paper Title | Method | Merits | Demerits | Paper Link |
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| Predicting the energy output of wind turbine based on weather condition | Predicted the output power of the wind turbines using the random forest regressor algorithm. The wind direction, wind speed and outdoor temperature are used as input parameters to predict output power | Low over-fitting tendency Simple and fast to train Low prediction error | Problem of missing data is not solved Prediction model does not predict other parameters like fault in wind turbine | https://ieeexplore.ieee.org/stamp/stamp.jsp?tp=&arnumber=9208852&tag=1 |
| Short term wind speed forecasting for wind turbine applications using linear prediction method | Utilizes the 'linear prediction' method in conjunction with 'filtering' of the wind speed waveform. The filtering eliminates the undesired parts of the frequency spectrum (i.e. smoothing) of the measured wind speed which is less effective in an application | Fits a linear differential equation to the data waveform Performs an accurate modelling high correlation between the output of the method and the real wind speed data | Increasing the model order can cause instability of obtained model Filtering out the less effective frequency components from the wind speed spectrum is not done | https://reader.elsevier.com/reader/sd/pii/S0960148107000237?token=E285B3E00332AD147BDFFE8E0BAD30E5B0E8D860E1F2BB81DA2870A89B488B9E7F6A7164501C2894664C042B93CE763D&originRegion=eu-west-1&originCreation=20220910183934 |
| Wind power forecasting of an offshore wind turbine based on high frequency SCADA data and deep learning neural network | A deep learning neural network was constructed to predict wind power based on a very high-frequency SCADA database. Input features were engineered based on the physical process of offshore wind turbines, while their linear and non-linear correlations were further investigated | Investigated non-linear correlations Proposed approach can reduce the computational cost Retains high accuracy | Learning rate is high Requires very large amount of data in order to perform better | https://reader.elsevier.com/reader/sd/pii/S0360544220308008?token=2C188EDEC70600F84640AAFD559E3E6B4A953B312803DDAC762A953E86506642189F88C4FB1FC4E10ACABBB34C870F2C&originRegion=eu-west-1&originCreation=20220910184930 |
| Wind turbine power output prediction model design based on artificial neural networks and climatic spatiotemporal data | Building a prediction model using the Artificial Neural Networks, activation function, analyze model performance for different sites, comparison on different climatic conditions | Model performance is compared across various sited to improve accuracy It can handle large amount of data sets | Doesn't take additional climatic variables like atmospheric pressure Data is not unified | https://ieeexplore.ieee.org/document/8352329?denied= |

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| Wind Turbine Power Output Estimation with Probabilistic Power Curves | Deterministic and probabilistic power curve, uses field data, Normal distribution and Weibull distribution are used to represent the probability density function of power output at various wind speed, Monte Carlo simulation is used to generate random predicting power output | Performs better than other deterministic models and probabilistic models Improves wind turbine power output estimation accuracy | Does not consider every wind turbine in the wind farm Need to extend model to calculate total wind power output | https://ieeexplore.ieee.org/stamp/stamp.jsp?tp=&arnumber=9209346 |
| A Review of Wind Power Forecasting Models | Wind power forecasting models for wind prediction, physical approaches, statistical approaches, adaptive fuzzy neural networks, on-line adaptation capabilities for optimal performance | presents a detailed review on existing tools used in wind speed wind power prediction over time-scales | Difficult to evaluate the performance of various models No forecasting model of can perfect any condition | https://www.sciencedirect.com/science/article/pii/S1876610211019291 |
| Wind Power Forecasts Using Gaussian Processes and Numerical Weather Prediction | Combination of numeric and probabilistic models: a Gaussian process (GP) combined with a numerical weather prediction (NWP) model, validated with three real-world datasets for model training and testing | Three real-world datasets were used for model training and testing The proposed model has improvement of accuracy for the regular large datasets | Model can not handle sparse dataset | https://ieeexplore.ieee.org/stamp/stamp.jsp?tp=&arnumber=6617679 |
| Wind turbine power output prediction using a new hybrid neuro-evolutionary method | K means clustering, SCADA time series decomposed by Hybrid Variational Mode Decomposition, which consists of VMD, GNM and ARLS heuristics, SaDE with sine cosine optimization hyper parameter tuning, LSTM | Model provides accurate forecasting Lowers computational runtime | Different results were achieved over iteration | https://www.sciencedirect.com/science/article/abs/pii/S0360544221008665 |