



# Data Analytics Application Series I: Wind Energy Part I

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# Why wind energy?

- Renewable energy plays an important role in the city sustainability
- Wind energy is one of most successfully commercialized renewable energy in the world
- Wind farm is composed of a population of distributed units, difficult to manage
- Wind turbine is working under harsh environment
- Wind is uncertain
- Wind energy asset is expensive
- Wind turbine is designed and manufactured to be fully instrumented

# A Commercial Wind Farm

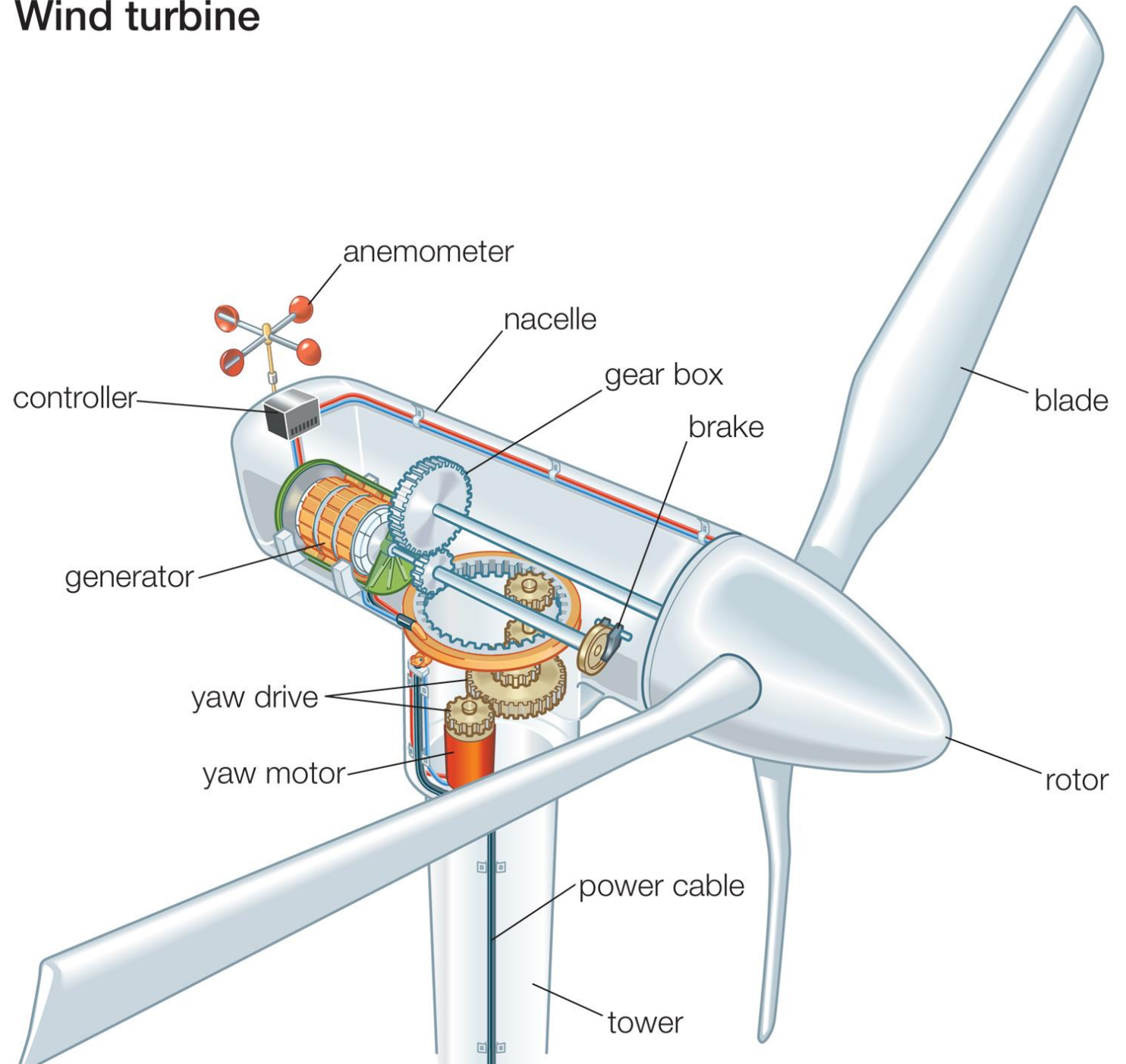




## Wind turbine

# A Wind Turbine System

- Composed of a number of mechanical and electrical components
- Energy conversion process is highly nonlinear
- A high dimension of system parameters
- Uncertain wind speed as free fuel

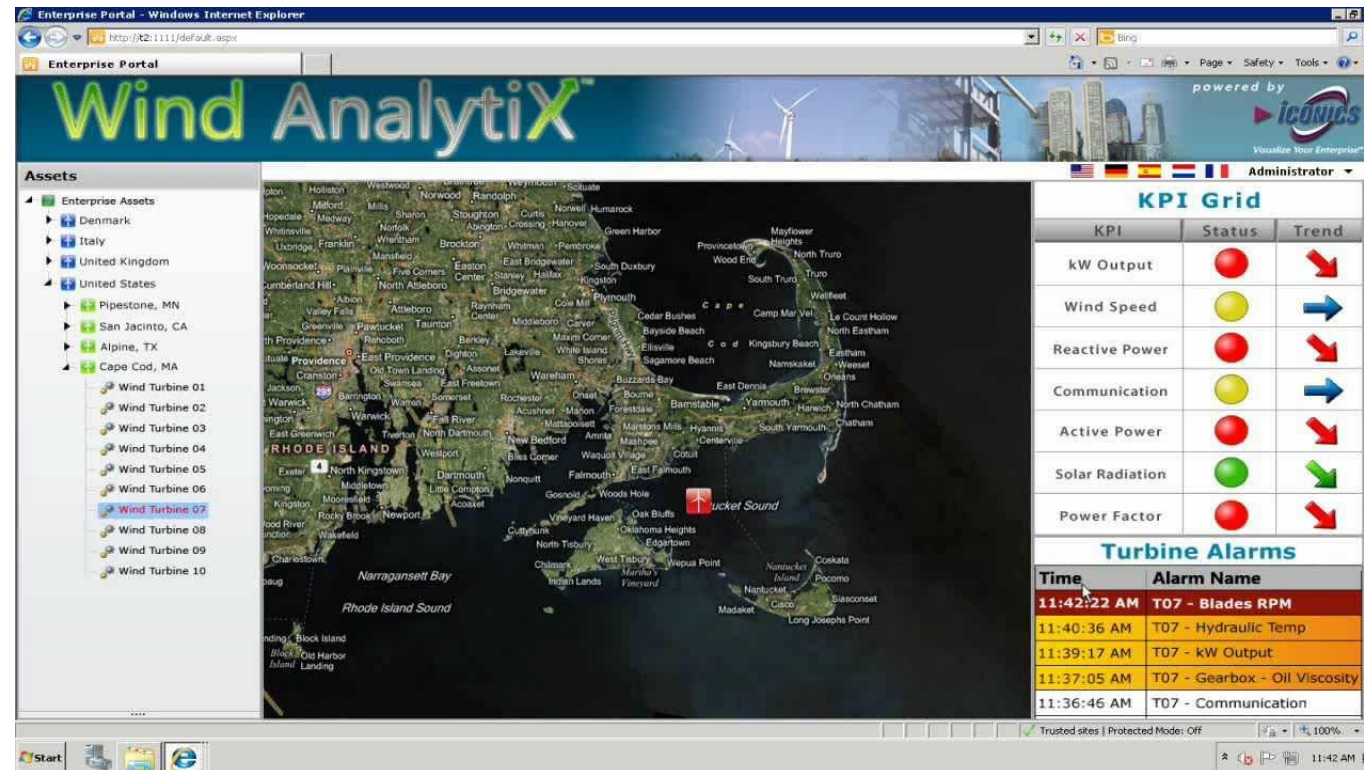


# Supervisory Control and Data Acquisition

- SCADA system, designed for monitoring WT conditions but it collects a massive volume of WT data
- Over hundred system and environmental parameters involved in wind power generation processes are measured
- Sampling frequency can be quite high but data typically stored in 10-min intervals due to the storage constraint
- SCADA data enables the era of wind energy big data analytics

# SCADA System

Can be considered as an IoT technology



# Wind Energy Problems Receiving Data Driven Solutions

## Problems Only Involve Machine Learning:

- Uncertain factor forecasting, such as wind speed, wind power.
- Data-driven wind turbine modeling

## Problem Involve Machine Learning + Another Field:

- Data-driven wind turbine intelligent control
- Data-driven wind farm operations
- Data-driven wind turbine condition monitoring

# Wind Energy ML Problems

- Wind speed/power prediction v.s. wind system modeling
- Problem formulation:
  - A data-driven modeling process considering physics based and mechanical principles
  - Based on domain knowledge, you need to know system inputs and system outputs
  - Input parameters – controllable and non-controllable
  - Output parameters – numerical or categorical based on the definition

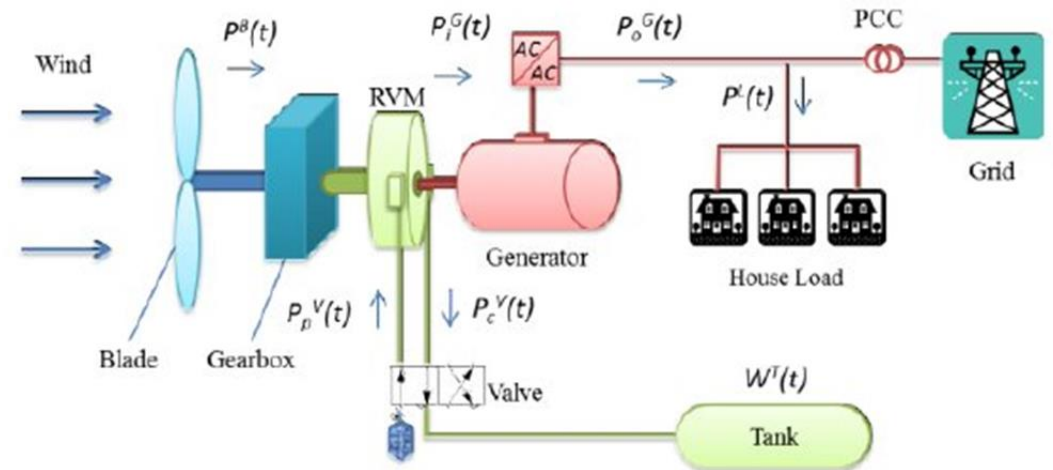
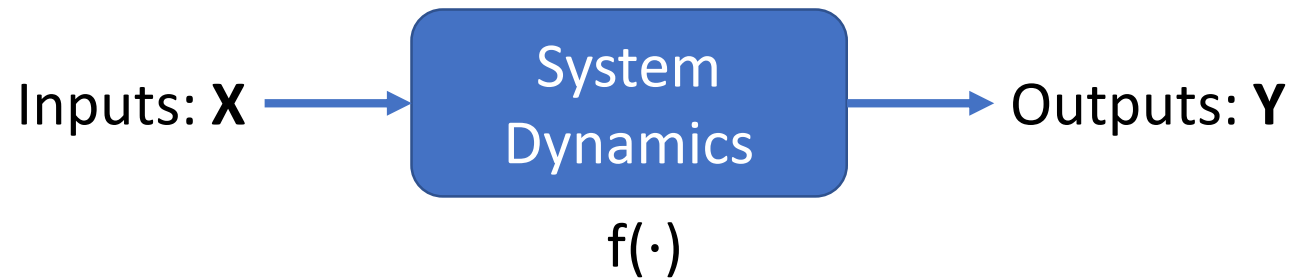


Figure 2. Proposed wind conversion system.



# Wind Energy ML Problem

A System



Problem:  $\mathbf{X}$ ,  $\mathbf{Y}$  are known as data, while  $f(\cdot)$  needs to be obtained

Given a loss function  $L$ , a data-driven system modeling task typically aims at identifying a non-parametric model  $g_A(\cdot)$  via algorithm  $A$  by minimizing the  $L(g_A(\mathbf{X}), \mathbf{Y})$ .

# Wind Energy ML Problem

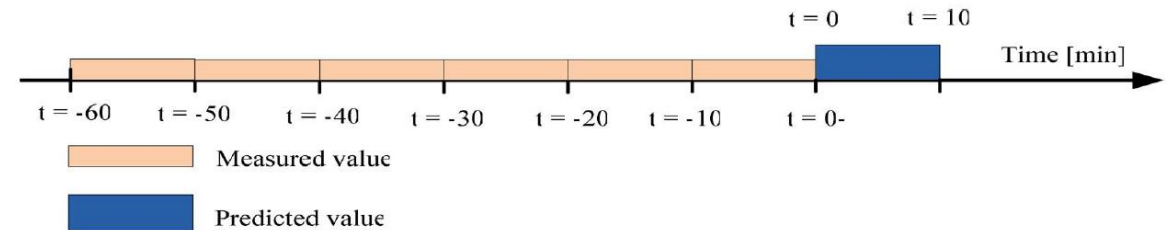
- Two types of problems can be generalized under the same computational framework
- Differences typically include the considered input parameters and the output parameters
- In modeling, we aim to use  $X_t$  to estimate  $Y_t$
- In prediction, we aim to use  $X_t$  to estimate  $Y_{t+n}$ ,  $n \geq 1$
- The learning process is the same, which aims to find a  $f(\cdot)$  correctly describing the relationship between  $X$  and  $Y$  based on data

# Wind Speed/Power Prediction

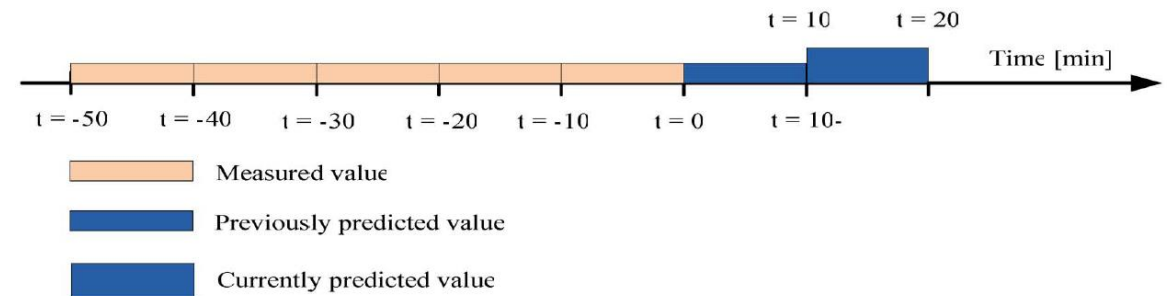
Source: A. Kusiak, H. Zheng, and Z. Song, "Short-Term Prediction of Wind Farm Power: A Data Mining Approach," IEEE Trans on Energy Conversion, Vol. 24, No. 1, pp. 125-136, 2009

Time-series model formulation

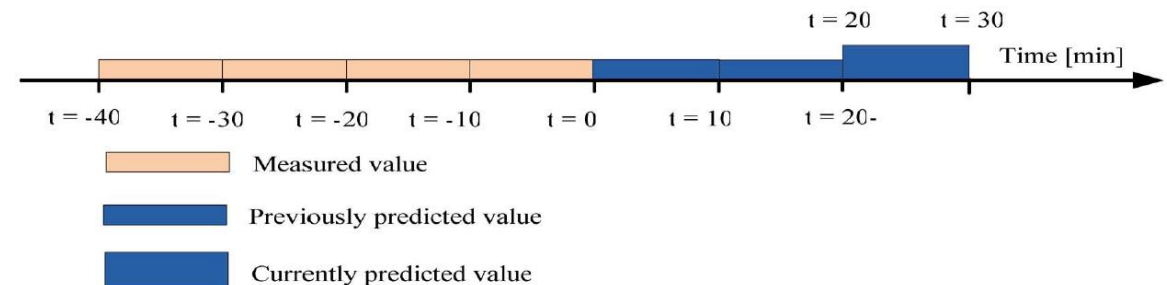
$$\hat{y}(t + T) = f(y(t), y(t - T), \dots, y(t - mT))$$



(a)



(b)



# Wind Speed/Power Prediction

Problems:

How to select appropriate input parameters (features) for doing data-driven modeling?

How to select the appropriate algorithm for developing/learning the data-driven models?



# Wind Speed/Power Prediction

## Feature selection

- Based on domain knowledge/principles
- Trial-and-error methods
- Linear models with feature selection capability, LASSO
- Nonlinear case
  - boosting tree – importance rank
  - NN – global sensitivity analysis
- No theoretical guarantee of the global optimal choice due to two reasons: 1) features measured might be insufficient; 2) feature selection and modeling algorithms can be different

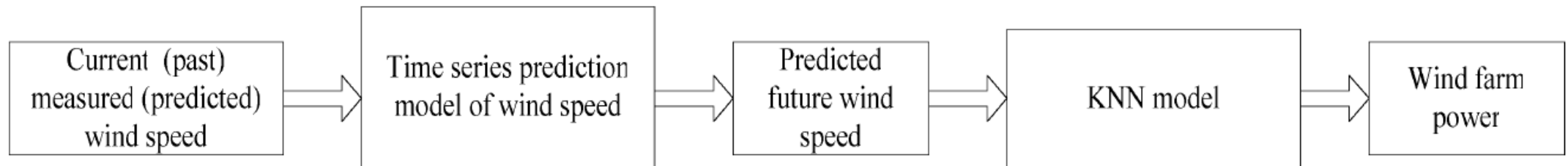
# Wind Speed/Power Prediction

## Algorithm selection

- A pool of candidate algorithms need to be determined
- Computational experiments for comparative analytics
- Assessment metrics, MAE, MSE, MAPE, and standard deviations
- All computational should be conducted on training dataset
- Training dataset can be split into training and validation

# Wind Speed/Power Prediction

- Ideas can be more inspiring

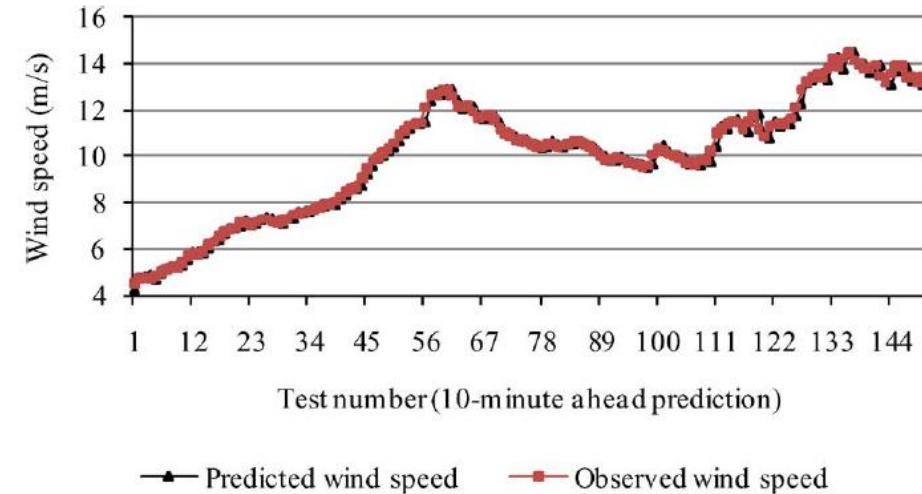


# Wind Speed/Power Prediction

- Illustration of results – Tables and Figures

TABLE III  
ERROR STATISTICS OF DIFFERENT MODELS BASED ON DATASET 3 OF TABLE I

Algorithm	Mean Absolute Error (m/s)	Absolute Error Std (m/s)	Mean Relative Error (%)	Relative Error Std (%)
SVMreg	0.198	0.181	3.514	5.730
MLP	0.266	0.388	4.661	6.739
M5P tree	0.206	0.267	3.953	6.035
REP tree	0.249	0.256	4.576	6.564
Bagging tree	0.202	0.198	3.665	5.723





# Wind Speed/Power Outputs

- kNN based integrated prediction model
  - For new data coming in, divide it into X and Y
  - Compute the distance of the new data to past records in the whole dataset based on X
  - Find K nearest historical records close to X
  - Average their predictions of ys to make it as a prediction of y of the new data, or stack a new regression model on top of it

# Wind Speed/Power Prediction

- Time series models only consider past records of the same parameter
- E.g.,  $X = \{y_{t-1}, y_{t-2}, \dots, y_{t-m}\}$  in the paper
- It is also to incorporate exogenous inputs, but must be records of them in the past as well