

B.Selection of input variables

- Input variables have to be selected keeping these 3 aspects in mind:
 - Which exogenous Variables have significant correlation to wind generation
 - Which station's data is suitable for our forecasting model
 - How many lagged hours to be included in the wind generation time series
- Table 1 shows that wind generation has positive cross correlation to 2 exogenous variables namely wind direction and gust speed.
- Table 2 shows the cross correlation between different stations and wind generation. We set a benchmark of 0.60 and select stations which have a correlation (>0.60) i.e. Stations 2 , 3 , 5.

TABLE I

CROSS-CORRELATIONS BETWEEN AGGREGATED WIND GENERATION AND DIFFERENT OBSERVATIONS WITHIN THE WIND FARM

	Wind direction	Barometric pressure	Temperature	Humidity	Gust speed
Aggregated WG	0.14	-0.02	-0.02	-0.07	0.62

TABLE II

CROSS-CORRELATIONS BETWEEN THE WIND SPEEDS IN THE WEATHER STATIONS AND THE AGGREGATED GENERATION OF THE WIND FARM

Stations	1	2	3	4	5	6	7
Cross-correlation With generation	0.56	0.66	0.67	0.54	0.62	0.50	0.51

TABLE IV
LIST OF INPUT DATA OF THE BCD CLASSIFIER

Input	Variable	Detail description
1-6	G_I	Wind power generation series
7-30	S_I	Wind speed series
31-40	Variables to determine prior probability	G_0 : Average wind generation
		S_0 : Average wind speed
		D : Forecasted wind direction
		H : Average humidity

- Table 3 is used for deciding the input variables for the SVR network.
- This table consists of hourly wind generation series of previous 6 hours as well forecasted and observed wind speeds of wind farm and weather stations 2, 3, 5 that were chosen.
- Here lag 1-6 indicates the observed wind generation for the previous 6 hours
- Similarly for wind speeds lag 0 indicated the forecasted wind speed at and lag 1-5 are for the observed wind speeds for previous 5 hrs

- Table 4 is used for deciding the input variables for the BCD classifier.
- First 6 input variables - Wind power generation time series.
- Next 24 input variables - Wind speed time series.
- Last 10 input variables - Avg wind generation , Avg wind speed , Forecasted wind direction and average humidity which is used to determine the prior probability of the classifier.

TABLE III
LIST OF INPUT DATA OF THE SVR NETWORK

Input	Variable name	Lagged value (hours)
1-6	Hourly wind power generation (G_I)	1,2,3,4,5,6
7-12	Hourly wind speed 0 (S_I)	0,1,2,3,4,5
13-18	Hourly wind speed 2	0,1,2,3,4,5
19-24	Hourly wind speed 3	0,1,2,3,4,5
25-30	Hourly wind speed 5	0,1,2,3,4,5

The wind speed 0 means the speed within the wind farm, and accordingly number 2, 3, or 5 means weather station 2, 3, or 5.

The hour of the predication is assumed at 0, the lag 0 represents the target instant, and the 6 lagged hours means the values that were measured 6 h earlier than the hour of predication.

C.The learning Algorithm - BCD classifier

- Method of clustering is used in the classifier using the principle of similarity of time series
- Step 1 of BCD classifier is to model the input time series using an autoregressive model. The matrix form of the model is : $x_j = X_j \beta_j + \varepsilon_j$ (eqn 4). Using given data and standard bayesian techniques , parameters are calculated.
- Step 2 of the BCD classifier is the forming of initial clusters which is done using a model based bayesian procedure. The matrix form of these clusters is $x_k = X_k \beta_k + \varepsilon_k$ (eqn 5). Here $x_k = \begin{pmatrix} x_{k1} \\ \vdots \\ x_{km_k} \end{pmatrix}$, $X_k = \begin{pmatrix} X_{k1} \\ \vdots \\ X_{km_k} \end{pmatrix}$.
- Step 3 of the BCD Classifier is ranking these initial clusters using the posterior probability given by $P(M_C | x) \propto P(M_C) f(x | M_C)$ (eqn 6). To calculate the marginal likelihood we use $f(x | M_C) = \int f(x | \theta_C) f(\theta_C) d\theta_C$ (eqn 7) which is simplified to remove the integral and results in $f(x | M_C) = \frac{\Gamma(1)}{\Gamma(1+m)} \times \prod_{k=1}^c \frac{\Gamma(m_k/m + m_k)}{\Gamma(m_k/m)}$ (eqn 8).

$$\times \frac{(RSS_k/2)^{(q-n_k)/2} \Gamma(n_k - q)/2}{(2\pi)^{(q-n_k)/2} \det(X_k^T X_k)^{(1/2)}}$$
- However these clusters grow exponentially and Step 4 of the BCD classifier is to reduce the no of clusters by merging them. There are two ways to do this :-
 - Agglomerative search strategy - Uses marginal likelihood to merge clusters
 - Heuristic strategy applied on agglomerative clustering - Uses a similarity measure namely euclidean distance between 2 times series $S_i = \{x_{i1}, \dots, x_{in}\}$ and $S_j = \{x_{j1}, \dots, x_{jn}\}$ given by $D(S_i, S_j) = \sqrt{\sum_{t=1}^n (x_{it} - x_{jt})^2}$. (eqn 9)