Variable importance analysis can increase the accuracies of the models

The variable importance measure must be weighted sums of the absolute regression coefficients.

The proposed prediction system will iteratively test the model with training and testing datasets

Cross-validation can used to evaluate method for reducing scales of overfitting and increasing accuracy of the model

Parameters like temperature, turbidity, pH and dissolved solids can be used Network structure selection method is proposed to identify the corelated input parameters

Massive dataset
and strong
correlation
between
parameters will
make the best
prediction.

Prediction can also be taken from the historical dataset

Use a minimal number of parameters with cheap sensors to predict water quality

Accurate model can be selected based on the outcome in the model evaluation

Stratified sampling strategy is used to mitigate the uneven distribution of training and testing dataset

Evaluating the effect of substantial nutrient loads on overall water quality

Keep the data design

The timeline of the measurements must be recorded

The data
distribution in the
testing data
should not affect
the training data
set.

Data modeling to use the past dataset to inform the future effort

Each data
needs to be in
different
measures to
analyze the
quality

The size of training datasets should not be less than the number of training parameters required in the model.

Feature selection
helps to simplify
the procedure and
reduce
computational
cost of analysis

Some of the variables can be eliminated due to the meaningless analysis

A method like neurofuzzy interference system can be implemented which is capable of integrating linear and non-linear relationships in dataset. The data mining techniques will be used for applying the classification method for water quality application

Using supervised
learning
algorithm, water
quality class can
be predicted

Various
techniques can
be included to
predict the
quality within
the application.