

Predicting CO2 Emission Using Machine Learning

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ABSTRACT

The air quality monitoring system collects data of pollutants from different location to maintain optimum air quality. In the current situation, it is the critical concern,. The introduction of hazardous gases into the atmosphere from industrial sources, vehicle emissions, etc. pollutes the air. Today, the amount of air pollution in large cities has surpassed the government-set air quality index value and reached dangerous levels. It has asignificant effect on a human health. The prediction of air pollution can be done by the Machine Learning (ML) algorithms. Machine Learning (ML) combines statistics and computer science to maximize the prediction power. ML is used in order to calculate the Air Quality Index. Various sensors and an Arduino Uno microcontroller are utilized to collect the dataset. Then by using K- Nearest Neighbor (KNN) algorithm, the air quality is predicted.

Keywords: Machine Learning, KNN,

1. INTRODUCTION

Among the most crucial challenges faced in the world today is air pollution. Industrial activity is increasing more regularly due to the explosive growth of economy, which is causing air pollution to increase more rapidly. Environmental pollution is a serious issue that affects all living things, including humans, with pollution from industry accounting for a significant portion of it. Solid particles such as dust, pollen, and spores, and gases, contribute to air pollution. Carbon monoxide, Carbon dioxide, Nitrogen dioxide, Sulphur oxide, Chlorofluorocarbons, Particulate Matter, and other air pollutants that cause air pollution are released by the combustion of natural gas, coal, and wood, as well as factories, cars, and other sources. Prolonged exposure to air pollution leads to serious health problems, such as lung and respiratory illnesses

The annual death toll from household exposure to gasoline smoke is 3.8 million. Exposure to the outdoor air pollution will cause 4.2 million deaths annually. 9 out of 10 people on the earth reside in areas with air quality that is worse than recommended by the World Health Organization. As per the Greenpeace Southeast Asia Analysis of IQAir statistics, air pollution and associated problems caused over 120,000 deaths in India in 2020. According to the report, air pollution caused economic losses of ₹2 lakh crore in India. This demonstrates how crucial it is to pay attention on the air quality.

Primary pollutants and the secondary pollutants are the two major classifications of air pollutants. One that is directly emitted into the atmosphere from its source is

The program for collecting the information is written in

referred to as a primary pollutant, whereas a secondary pollutant is one that is produced due to the interaction between two primary pollutants or with other elements of the atmosphere. One of the detrimental effects of pollutants emitted into the environment is the degradation of air quality. Also, other harmful effects, such as acid rain, global warming, aerosol production, and photochemical smog has increased in past years.

Predicting the air quality is crucial for preventing the problem of air pollution. The Machine Learning (ML) models can be used for this. With the use of training data, a computer can learn how to build models via a technique called as Machine Learning. It is a branch of Artificial Intelligence that gives computer program the ability to forecast outcomes with ever-increasing accuracy. ML can examine a variety of data and identify patterns and particular trends. Machine learning is the ability given to a computer program to do a task without any external programming and this is task is achieved by using some statistical and advanced mathematical algorithms.

As air pollution has been rising every day, monitoring has proven to be a significant task. The amount of pollution in a given area is determined through continuous air quality monitoring at that location. The information obtained by the sensors reveals the source and concentration of the pollutants in that area. Measures to minimise pollution levels can be taken usingthat knowledge and the ML model.

The hardware device consists of three different sensors like MQ-135 air quality sensor, MQ-5 sensor, Optical dustsensor connected to the Arduino uno board, which helps in collecting the pollutants information of the currentplace.

the Arduino IDE according to the AQI level specified by the Central Pollution Control Board of India, in the report



National Air Quality Index. The information collected from the sensors is recorded in the excel sheet, then it is stored in the required file path, which makes the dataset. Further the excel sheet in the .csv file is directly read in the ML program.

2. LITERATURE SURVEY

The authors of [1] proposed that Machine Learning algorithms plays important role in measuring air quality index accurately. Logistic regression and auto regression, ANN help in determining the level of PM2.5. ANN comes out with best results in the paper.

In [2] authors gives the prediction of the air quality indexby using different machine learning algorithms like Decision Tree and Random Forest. From the results, concluded that the Random Forest algorithm gives better prediction of air quality index.

In [3] authors proposed model by using BILSTM which is the Deep Learning model to predicted the PM2.5 with improved performance comparing the existing model and produced exceptional MAE, RMSE.

In [4] authors used the prediction model results were based on Big Data Analytics and Machine Learning, which have helped to evaluate and contrast current assessments of air quality. The Decision Tree algorithm gave the best results among all the algorithms.

The authors of [5] used SVR, and LSTM Machine Learning models. The Machine Learning algorithms used for estimating the atmospheric pollutants (PM10 and PM2.5), it was demonstrated that SVR algorithms are themost suitable in forecasting the air pollutants concentrations.

Various studies have contributed significantly to the field of CO2 emission prediction models using diverse methodologies. Saleh et al. (2016) employed Support Vector Machines (SVM) to forecast CO2 emissions, emphasizing environmental management applications. Garip and Oktay (2018) applied machine learning techniques to predict CO2 advancing emissions, climate change forecasting capabilities. Shih and Tsokos (2008) explored prediction models for carbon dioxide emissions and their atmospheric impacts, providing foundational insights. Amarpuri et al. (2019) introduced a hybrid deep learning approach for CO2 emission prediction, particularly effective in the Indian context. Vijayumar and Kadam (2018) developed a machine learning-based model for CO2 emissions, demonstrating practical applications. Yeasmin et al. (2020) integrated artificial intelligence for CO2 emission predictive analysis, showcasing technological advancements in environmental monitoring. Additionally, Ho et al. (2015) focused on sector-specific prediction models for CO2 emissions in Malaysia's manufacturing and construction industries, addressing regional environmental challenges. Kangralkar

and Khanai (2021) applied machine learning to predict automotive emissions, contributing to sustainable transportation solutions. These studies collectively highlight the diversity of approaches and applications in CO2 emission prediction research, spanning machine learning, deep learning, and sector-specific modeling strategies.

3. METHODOLOGY

Information about air pollutants is obtained from the sensors, analysed, and then saved as a dataset. This dataset has been pre-processed with a variety of features, which includes attribute selection and normalisation. Once it is available, the dataset is divided into a training set and a test dataset. The training dataset is then used to apply a Machine Learning algorithm. The obtained results are matched with the testing datasetand results are analysed.

3.1 Machine Learning model

Machine Learning algorithm is implemented to predict the air pollution. Machine Learning (ML) is a subfield of Artificial Intelligence (AI) that enables the software applications to be accurate in predicting theoutcomes without being explicitly programmed to do so.

To predict the new outcomes, Machine Learning algorithms make use of existing past data as the input. With the help of Machine Learning, a user can provide a computer program huge amount of data, and the computer will only examine that data and draw conclusions from it.

KNN is the Machine Learning algorithm used for the prediction of air pollution. The K-Nearest Neighbors (KNN) algorithm is one of the types of Supervised Machine Learning algorithms. KNN is incredibly simple to design but performs quite difficult classification jobs. KNN is called the lazy learning algorithm as it lacks the training phase. Instead, it classifies a fresh data point while training on the entire dataset. It does not make any assumptions, hence it is called non-parametriclearning method.

Steps in KNN:

- Determine the distance between each sample of the training data and the test data.
- To determine distance, we can utilise the Euclidian or Minkowski or Manhattan distance formula.
- Sort the estimated distances in ascending order.
- Vote for the classes.
- Output will be determined based on class havingmost votes.
- Calculate the Accuracy of the model, if required rebuild model.



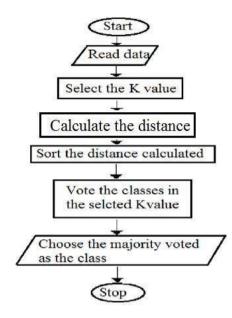


Fig-1: Flow chart of KNN

3.2 Air Quality Index

AQI	Associated Health Impacts
Good (0-50)	Minimal Impact
Satisfactory (51–100)	May cause minor breathing discomfort to sensitive people
Moderate (101–200)	May cause breathing discomfort to the people with lung disease such as asthma and discomfort to people with heart disease, children and older adults
Poor (201–300)	May cause breathing discomfort to people on prolonged exposure and discomfort to people with heart disease with short exposure
Very Poor (301-400)	May cause respiratory illness to the people on prolonged exposure. Effect may be more pronounced in people with lung and heart diseases
Severe (401-500)	May cause respiratory effects even on healthy people and serious health impacts on people with lung/heart diseases. The health impacts may be experienced even during light physical activity



The Central Pollution Control Board of India provided the AQI in the report National Air Quality Index, which is shown in the Fig 2 above.

According to this AQI the program is written in the Arduino IDE to collect the dataset from the current place. Further this dataset is recorded in the excel sheet and saved in the particular file path as required.

3.3 ARCHITECTURE

Data Input: The Input dataset is provided to the models in the form of a .csv file. Data Processing: In the .csv file if there are any corrupted, incorrectly formatted, duplicated, or incomplete datasets then it is removed and then this dataset is organized to look similar across all fields. Algorithm Training: In this process of training an ML model involves providing an ML algorithm (that is, Random Forest and Support Vector Machine) with 60% training data to learn from. The learning algorithm finds patterns in the training data that map the input data

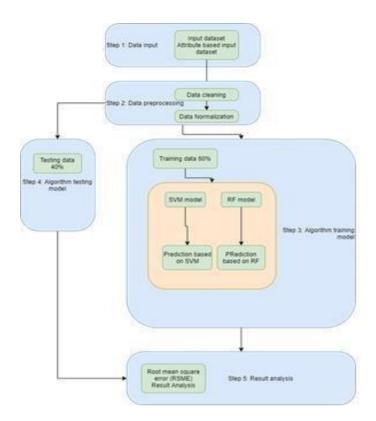


Fig-3: Block diagram

attributes to the target (the answer that you want to predict), and it outputs an ML model that captures these patterns.

Algorithm Testing: It is the actual execution of the algorithm. Where the User provides input (that is, 40% of the dataset) and the algorithm calculates the output based on the learned parameters from the training phase. Result Analysis: In this step, the output is analyzed on the bases of MAE, MSE, RMSE, and R2- scores.

The software used is Anaconda Navigator for Python, which features the web-based Interactive Development Environment for data, code and notebooks called Jupyter Notebook. The users could build and arrange the workflows in data sciences, machine learning and scientific computing, using its interface. The Jupyter Notebook is the original web tool for producing computational documents.



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Fig-4: Data collection

Steps in Software Implementation

1) Read dataset:

The required libraries are imported and then the datasetis read in the python code.

Parameters considered in the dataset are the sensor values of air quality, smoke and dust and the respective quality of air for the values from the sensors of the current place. Hence, there are four columns in the dataset and the number of rows depends on the time that the data is recorded. This dataset is saved as the .csvfile in excel.

	air	smoke	dust	quality	
0	61	37	50	satisfactory	
1	61	37	50	satisfactory	
0 1 2 3	61	37	50	satisfactory	
3	61	37	50	satisfactory	
4	61	37	50	satisfactory	
1114	63	36	498	severe	
1115	63	36	498	severe	
1116	63	36	498	severe	
1117	63	37	498	severe	
1118	63	36	498	severe	

Fig-5: Reading dataset

2) Split the training and testing dataset:

The training set is used to train the model, and thetesting set is used to determine whether the model generalises well to new and unexplored data. The better outcomes are attained when 20% to 30% of the data are used for testing and the rest 70% to 80% for training.

This is done by importing the train_test_split library from the Sci-kit, where training and testing ratio is taken 80% and 20% respectively.

3) Choosing the Machine Learning model

plt.scatter(y_test,KNN_predicted)

KNN is the Machine Learning model chose for the prediction of air pollution.

4) Prediction

After the ML model is fit, it gives the prediction of air quality based the AQI described, of the current place i.e., whether the air quality is satisfactory or moderate tobreathe, or poor so that people can decide the impact of air pollution, or very poor and severe to survive in that place.

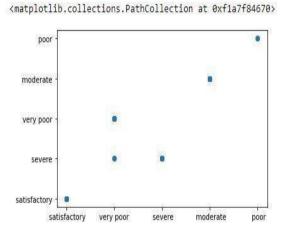


Fig-6: Scatterplot of y_test and predicted values

4. RESULT AND DISCUSSION

The confusion matrix of the particular dataset air pollution, which is read initially at the time of read dataset is as shown in the Fig 8 below.

confusion				A COLUMN T		ion_matrix ted)
array([[17,	0,	0,	0,	0],	
]	0,	2,	0,	0,	0],	
[0,	0,	92,	0,	0],	
[0,	0,	0,	102,	0],	
[0,	0,	0,	2,	9]],	dtype=int64)

Fig-7: Confusion matrix

The accuracy of confusion matrix shown in the figure is,

Accuracy = (17 + 2 + 92 + 102 + 9) /

$$(17 + 2 + 92 + 102 + 9 + 2)$$

= 222/ 224

= 99.1071 %

	HOME WEW DATA RANDOMFOREST SVR FORECAST LOGO
11	Forecast
	lower co2 upper co2
	2021-10-01-481366.823092 505235.875595
	2821-11-01-482667.869586 506036.916089
	2021-12-01-483467.614186 506836.660689
	2022-01-00-483910.522022 507724.693220
	2022-02-04-454723.377794 500537.548992
	2022-03-01-455534.905303 509349.075501
	2022-04-02-456345.111040 510159.202230
	2022-05-01-457154.001442 510968.172640
	2022-06-01-487901.582095 511775.754093
	2022-07-02-486767.861732 512562.032930
	2822-88-82-489572.844236 513387.015435
	2822-09-01-450376-536641 514190.707839

Fig-8: Predicted outcomes

5. CONCLUSION

In conclusion, the field of predicting CO2 emissions using machine learning has seen significant advancements and diverse applications across various domains. Studies reviewed here, ranging from Support Vector Machines (SVM) and deep learning hybrids to sector-specific models, illustrate the versatility and effectiveness of machine learning in forecasting CO2 emissions. These models not only contribute to environmental management and policymaking but also offer practical solutions for mitigating climate change impacts. Moving forward, continued research and innovation in machine learning techniques hold promise for further enhancing the accuracy and scalability of CO2 emission predictions, essential for achieving sustainable environmental practices and global climate goals.

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