

PREDICTION OF AIR QUALITY USING ML ALGORITHMS

¹JONNALAGADDA VIJAYA SAI SUDHEER, ²JONNALAGADDA GOPI CHAND,
³DIVI SUDHEER, ⁴TANIKONDA DANIEL RAJU, ⁵MS. FEMIMOL.R

^{1,2,3,4}students, ⁵assit Professor,
COMPUTER SCIENCE DEPARTMENT
Bharath Institute of Science &Technology affiliated to
Bharath Institute of Higher Education and Research,
Chennai, Tamil Nadu, India

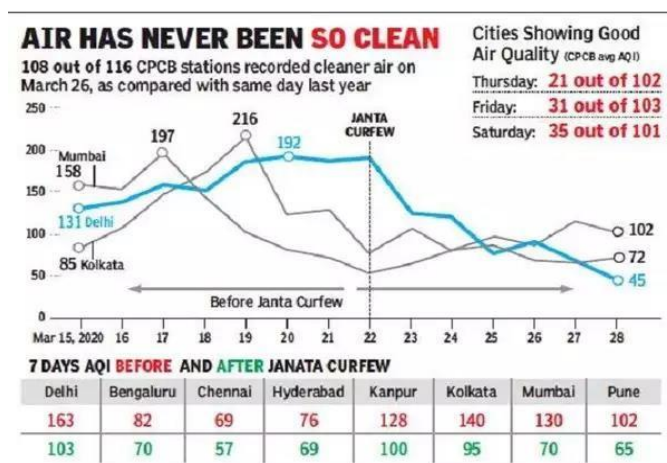
Abstract - Air pollution is a growing issue these days. It is necessary to monitor air quality and keep it under control for a better future and healthy living for all. Here we propose a prediction of air quality using ML. The system predicts air quality. Examining and protecting air quality has become one of the most essential activities for the government in many industrial and urban areas today. The meteorological and traffic factors, burning of fossil fuels, and industrial parameters play significant roles in air pollution. With this increasing air pollution. The deposition of this harmful gases in the air is affecting the quality of people's lives, especially in urban areas. Sulphur dioxide irritates the skin and mucous membranes of the eyes, nose, throat, and lungs. By using the machine learning regression models on AQI and classification models on AQI range, the system predicts the air quality in the respective locality.

Key Words: Air pollution, AQI range, machine learning

1. INTRODUCTION:

Nowadays the air condition is very polluted. In recent years, car emissions, chemicals from factories, smoke, and dust are everywhere. Air pollution cannot be detected by human feelings. Air pollution may contain a lot of dangerous substances such as ozone, particulate matter, Sulphur dioxide, nitrogen dioxide, carbon monoxide, and lead. Humans need an atmosphere of air that is free from contaminants. This is very crucial for human life and health. Any change in the natural composition of air may cause grave harm to life forms on Earth. The effect of Air Pollution ranges from difficulty in breathing, coughing etc., polluted air can also impair visibility. Air Pollution is one of the serious and major environmental problem worldwide. People are suffering from health problems as a result of prolonged exposure to polluted environments. Health problems have been growing at a faster rate especially in urban areas of developing countries where industrialization and growing number of vehicles leads to release of a lot of gaseous pollutants.

Fig:1.1 Air Quality Index Analysis report



Harmful effects of pollution include mild allergic reactions such as irritation of the throat, eyes and nose as well as some serious problems like bronchitis, heart diseases, pneumonia, lung and aggravated asthma. According to a survey, due to air pollution 50,000 to 100,000 premature deaths per year occur in the U.S. alone. Whereas in EU number reaches to 300,000 and over 3,000,000 worldwide. NASA on 19th November 2021 released a report upon the issue of pollution in Delhi.

According to the news, in the year 2019, 16.7 lakh people died due to polluted air in India. This is a big Disaster, which the administration is trying to fix, but this work is not looking easy. The number of vehicles increased rapidly as soon as the lockdown lifted. Continued exposure to environments with poor air quality is a major public health concern in developed and developing countries. It is estimated that the pollutants responsible for poor air quality cause nearly 2.5 million premature deaths per year worldwide. Significantly, around 1.5 million of these deaths are due to polluted indoor air, and it is suggested that poor indoor air quality may pose a significant health risk to more than half of the world's population.

According to the World Health Organization (WHO), each year air pollution is responsible for nearly seven million deaths around the globe. Nine out of ten human beings currently breathe air that exceeds the WHO’s guideline limits for pollutants, with those living in low- and middle-income countries suffering the most. Smog can irritate the eyes and throat and also damage the lungs, especially those of children, senior citizens, and people who work or exercise outdoors. It’s even worse for people who have asthma or allergies: these extra pollutants can intensify their symptoms and trigger asthma attacks. The tiniest airborne particles in soot, whether gaseous or solid, are especially dangerous because they can penetrate the lungs and bloodstream and worsen bronchitis, lead to heart attacks, and even hasten death. A number of air pollutants pose severe health risks and can sometimes be fatal even in amounts.

AQI	Remark	Color Code	Possible Health Impacts
0-50	Good	Green	Minimal impact
51-100	Satisfactory	Light Green	Minor breathing discomfort to sensitive people
101-200	Moderate	Yellow	Breathing discomfort to the people with lungs, asthma and heart diseases
201-300	Poor	Orange	Breathing discomfort to most people on prolonged exposure
301-400	Very Poor	Red	Respiratory illness on prolonged exposure
401-500	Severe	Dark Red	Affects healthy people and seriously impacts those with existing diseases

Fig:1.2 AQI Range Chart

Almost 200 of them are regulated by law; some of the most common are mercury, lead, dioxins, and benzene. Another category of toxic compounds, Polycyclic Aromatic Hydrocarbons (PAHs), are by-products of traffic exhaust and wildfire smoke. In large amounts they have been linked to eye and lung irritation, blood and liver issues, and even cancer. In one study, the children of mothers exposed to PAHs during pregnancy showed slower brain-processing speeds and more pronounced symptoms of ADHD. People experience a wide range of health effects from being exposed to air pollution. Effects can be broken down into short-term effects and long-term effects. Short-term effects, which are temporary, include illnesses such as pneumonia or bronchitis. They also include discomfort such as irritation to the nose, throat, eyes, or skin.

Air pollution can also cause headaches, dizziness, and nausea. Bad smells made by factories, garbage, or sewer systems are considered air pollution, too. These odors are less serious but still unpleasant. Long-term effects of air pollution can last for years or for an entire lifetime. They can even lead to a person’s death. Long-term health effects from air pollution include heart disease, lung cancer, and respiratory diseases such as emphysema. Air pollution can also cause long-term damage to people’s nerves, brain, kidneys, liver, and other organs. Some scientists suspect air pollutants cause birth defects. Nearly 2.5 million people die worldwide each year from the effects of outdoor or indoor air pollution.

Pollutant	Time	2005 levels	New 2021 levels
PM _{2.5} Particulate matter <2,5 microns	Annual	10	5
	24-hour	25	15
PM ₁₀ Particulate matter <10 microns	Annual	20	15
	24-hour	50	45
O ₃ Ozone	Peak season	-	60
	8-hour	100	100
NO ₂ Nitrogen dioxide	Annual	40	10
	24-hour	-	25
SO ₂ Sulfur dioxide	24-hour	20	40
CO Carbon monoxide	24-hour	-	4

Fig:1.3 Annual AQI report

Accordingly, epidemiology studies have documented adverse respiratory and cardiovascular effects for populations living in close proximity to major roadways. Studies have shown specific adverse health effects of being exposed to Particulate Matter (PM), including heart diseases, cancer risk and adverse birth outcomes. Children, pregnant women, elderly and people with existing health issues are some of the vulnerable population groups to adverse traffic-related air pollution. Studies have shown an association between traffic-related air pollution and reduced fetal growth, preterm birth and post-term low birth weight and susceptibility to

asthma. We cannot completely eradicate this problem but we may take several measures so that we can reduce air pollution that is caused by the vehicles. We Proposed this System so that it will help to monitor the air pollution that was caused by the Vehicles and to inform to higher authorities regarding to this. so that they will take appropriate measures.

We cannot completely eradicate this problem but we may take several measures so that we can reduce air pollution that is caused by the vehicles. We Proposed this System so that it will predict the air quality. Air pollution is a growing issue these days. It is necessary to monitor air quality and keep it under control for a better future and healthy living for all. Here we propose an air quality predicting system using machine learning.

2. LITERATURE SURVEY:

Tapiwa M, et.al. [1] machine-learning techniques such as Linear Regression (LR) models and Artificial Neural Networks (ANNs) to forecast or estimate air quality has been done before even though the focus has largely been on different pollutants and with different methodologies. Historical air quality data from the city of Johannesburg and the Vaal Triangle in South Africa was collected. The data was analysed and machine-learning techniques were applied to the data to generate prediction models for ground level pollution. The results obtained are encouraging with the prediction accuracy ranging from fair to good

A. Masih [2] The major techniques applied for pollutant concentration estimation or forecasting; and whether these techniques were based on Linear Regression, Neural Network, Support Vector Machine or Ensemble learning algorithms. The results obtained suggest that, machine learning techniques are mainly conducted in continent Europe and America. Further more a factorial analysis named multi-component analysis performed show that pollution estimation is generally performed by using ensemble learning and linear Regression based approaches, whereas, forecasting tasks tend to implement neural networks and support vector Machines based algorithm

C. Amuthadevi, et.al. [3] Development of air quality monitoring (AQM) models using different machine learning approaches (i) detection and classification of gases by implementing classifiers such as ANNs, k-Nearest Neighbors (KNN) algorithm, hybrid support vector machine (HSVM) etc., and (ii) estimation of gas molecule concentrations by using multivariable regression models such as linear regression, ANN, Deep learning Long-Short-Term Memory (DL-LSTM). Many air monitoring places were established for monitoring and collecting the air-pollution related data for research. By periodic monitoring, the attentions of air pollutants in subsequent hour otherwise the subsequent-day could be predicted. Based on these predictions, government can be aware of it and control vehicles at that location

Dr. D.J. Samatha, et.al. [4] ML models are able to achieve higher accuracies with large datasets, than classic statistical methods. Such models have long been used for AQI forecasting tasks. ML models are nonlinear, nonparametric in nature and hence are better able to handle the complexity of nonlinear elements like pollutant levels in the air. The results show that machine learning models (logistic regression and auto regression) can be efficiently used to detect the quality of air and predict the level of AQI in the future. The proposed system will help common people as well as those in the meteorological department to detect and predict pollution levels and take the necessary action in accordance with that.

Sriram Krishna Yarragunta, et.al. [5] The supervised machine learning technique (SMLT) was used to analyse a dataset and capture multiple pieces of information, including variable recognition, uni-variate analysis, bi-variate and multivariate analysis, missing value treatments, and data analysis. The agenda of our work is not only to bring awareness but also to minimize pollution through proper measures and ensure that the vehicles are emitting the pollutants within the range of regular pollution check. This can lead to a pollution free region in the area.

Elia Georgiana Dragomir. [6] Air quality index prediction using K-nearest neighbour technique Data mining techniques, as artificial neural networks, genetic algorithms, decision trees, k -nearest neighbour, logistic regression have been successfully used in air quality prediction problems. In this paper, we have presented an experiment done in order to determine the particularities of applying this technique for air quality analysis. Aiming at generating a prediction for the air quality index, training data that were collected in June 2009 were used as input data for the algorithm. The experimental results show that among the parameters that have been selected for this experiment, there is a strong correlation, and, therefore, these can be used in the forecasting processes

Mahalingam U, et.al. [7] A machine learning model for air quality prediction for smart cities. This paper addresses the challenge of predicting the Air Quality Index (AQI), with the aim to minimize the pollution before it gets adverse, using two Machine Learning Algorithms: Neural Networks and Support Vector Machines. The air pollution databases were extracted from the Central Pollution Control Board (CPCB), Ministry of Environment, Forest and Climate change, Government of India. The proposed Machine Learning (ML) model is promising in prediction context for the Delhi AQI. The results show improvement of the prediction accuracy and suggest that the model can be used in other smart cities as well.

Kingsy Grace R, et.al. [8] A comprehensive review of wireless sensor networks based air pollution monitoring systems. This paper presents a comparative study about the literature for air pollution monitoring systems based on the classification such as stationary air pollution monitoring systems, dynamic air pollution monitoring systems and pollution data analysis techniques. These pollution monitoring systems are compared based on the methodologies followed, microcontroller used, communication device used, pollutants analyzed using sensors, evaluation attributes, tested location and performance of the system.

Gallego E, et.al. [9] Outdoor air monitoring. Use of a semiconductor gas sensor for activating a sampler during pollution episodes. Simultaneous 24 h and episodic samples were taken during 15 days in El Morell. Higher levels of VOCs, not specific compounds, trigger the activation of the sampler. Sensor interferences derived from relative humidity were not observed. The study proposes an expansion of the field of use of semiconductor gas sensors. Hence, these aspects validate the use of the evaluated sensor for its application for the activation of samplers in air quality evaluations when episodic event occur, an interesting and innovative technique.

3. SYSTEM ARCHITECTURE

The following flow chart demonstrates the architecture of proposed system.

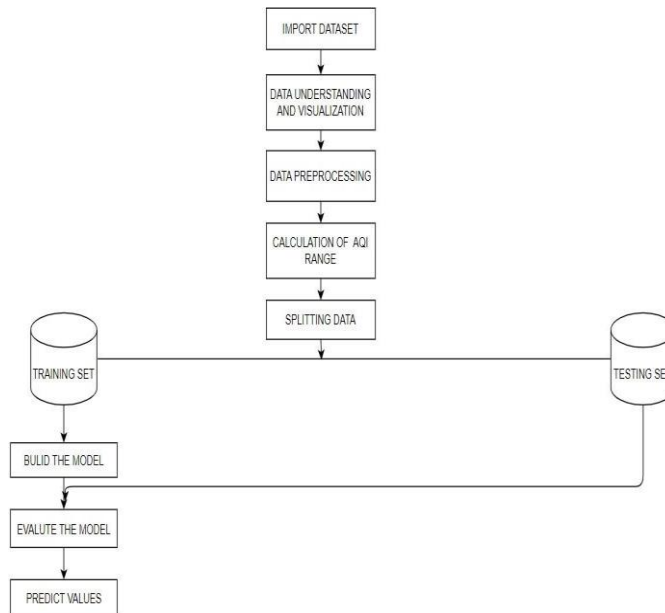


Fig :3.1 System Architecture

3.1 DATA SOURCE:

To predict the air quality index of a particular region ,we need the date, AQI, AQI Range
 As we taken the data from a particular location that is New-Delhi embassy road. The AQI formulae will be applied in order to calculate the AQI by using the KNN classification algorithm for a particular years. Several datasets will be imported inside the directory and null values will be set to the infinite data. The predicted and actual values will be represented using the Box-Plot analysis in order to remove the outliers.

The dataset consists of around 2843 records of the New-Delhi embassy road in India. This dataset consist of 3 attributes listed below.

- 1) Date
- 2) AQI
- 3) AQI RANGE

3.2 PREPROCESSING THE DATA:

A preliminary processing of data in order to prepare it for the primary processing or for further analysis.

Three steps in preprocessing of data for data cleaning 1.Missing value

2 Noisy Data

3 Removing unwanted data

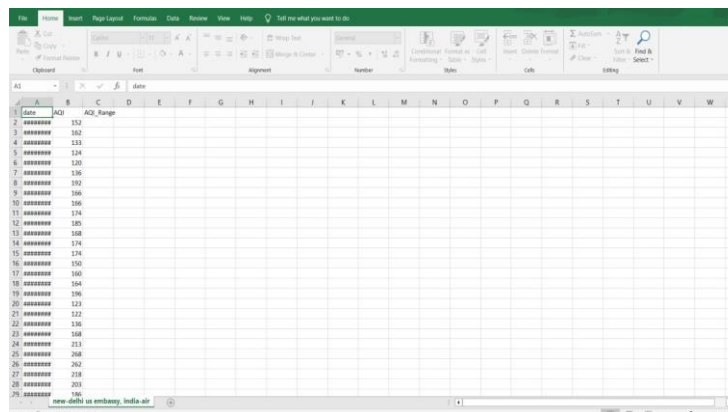


Fig :3.2 Before Pre-processing the Dataset

Date	AQI	AQI_Range
1	52	Poor
2	52	Poor
3	52	Poor
4	52	Poor
5	52	Poor
6	52	Poor
7	52	Poor
8	52	Poor
9	52	Poor
10	52	Poor
11	52	Poor
12	52	Poor
13	52	Poor
14	52	Poor
15	52	Poor
16	52	Poor
17	52	Poor
18	52	Poor
19	52	Poor
20	52	Poor
21	52	Poor
22	52	Poor
23	52	Poor
24	52	Poor
25	52	Poor
26	52	Poor
27	52	Poor
28	52	Poor
29	52	Poor
30	52	Poor
31	52	Poor
32	52	Poor
33	52	Poor
34	52	Poor
35	52	Poor
36	52	Poor
37	52	Poor
38	52	Poor
39	52	Poor
40	52	Poor
41	52	Poor
42	52	Poor
43	52	Poor
44	52	Poor
45	52	Poor
46	52	Poor
47	52	Poor
48	52	Poor
49	52	Poor
50	52	Poor
51	52	Poor
52	52	Poor
53	52	Poor
54	52	Poor
55	52	Poor
56	52	Poor
57	52	Poor
58	52	Poor
59	52	Poor
60	52	Poor
61	52	Poor
62	52	Poor
63	52	Poor
64	52	Poor
65	52	Poor
66	52	Poor
67	52	Poor
68	52	Poor
69	52	Poor
70	52	Poor
71	52	Poor
72	52	Poor
73	52	Poor
74	52	Poor
75	52	Poor
76	52	Poor
77	52	Poor
78	52	Poor
79	52	Poor
80	52	Poor
81	52	Poor
82	52	Poor
83	52	Poor
84	52	Poor
85	52	Poor
86	52	Poor
87	52	Poor
88	52	Poor
89	52	Poor
90	52	Poor
91	52	Poor
92	52	Poor
93	52	Poor
94	52	Poor
95	52	Poor
96	52	Poor
97	52	Poor
98	52	Poor
99	52	Poor
100	52	Poor
101	52	Poor
102	52	Poor
103	52	Poor
104	52	Poor
105	52	Poor
106	52	Poor
107	52	Poor
108	52	Poor
109	52	Poor
110	52	Poor
111	52	Poor
112	52	Poor
113	52	Poor
114	52	Poor
115	52	Poor
116	52	Poor
117	52	Poor
118	52	Poor
119	52	Poor
120	52	Poor
121	52	Poor
122	52	Poor
123	52	Poor
124	52	Poor
125	52	Poor
126	52	Poor
127	52	Poor
128	52	Poor
129	52	Poor
130	52	Poor
131	52	Poor
132	52	Poor
133	52	Poor
134	52	Poor
135	52	Poor
136	52	Poor
137	52	Poor
138	52	Poor
139	52	Poor
140	52	Poor
141	52	Poor
142	52	Poor
143	52	Poor
144	52	Poor
145	52	Poor
146	52	Poor
147	52	Poor
148	52	Poor
149	52	Poor
150	52	Poor
151	52	Poor
152	52	Poor
153	52	Poor
154	52	Poor
155	52	Poor
156	52	Poor
157	52	Poor
158	52	Poor
159	52	Poor
160	52	Poor
161	52	Poor
162	52	Poor
163	52	Poor
164	52	Poor
165	52	Poor
166	52	Poor
167	52	Poor
168	52	Poor
169	52	Poor
170	52	Poor
171	52	Poor
172	52	Poor
173	52	Poor
174	52	Poor
175	52	Poor
176	52	Poor
177	52	Poor
178	52	Poor
179	52	Poor
180	52	Poor
181	52	Poor
182	52	Poor
183	52	Poor
184	52	Poor
185	52	Poor
186	52	Poor
187	52	Poor
188	52	Poor
189	52	Poor
190	52	Poor
191	52	Poor
192	52	Poor
193	52	Poor
194	52	Poor
195	52	Poor
196	52	Poor
197	52	Poor
198	52	Poor
199	52	Poor
200	52	Poor
201	52	Poor
202	52	Poor
203	52	Poor
204	52	Poor
205	52	Poor
206	52	Poor
207	52	Poor
208	52	Poor
209	52	Poor
210	52	Poor
211	52	Poor
212	52	Poor
213	52	Poor
214	52	Poor
215	52	Poor
216	52	Poor
217	52	Poor
218	52	Poor
219	52	Poor
220	52	Poor
221	52	Poor
222	52	Poor
223	52	Poor
224	52	Poor
225	52	Poor
226	52	Poor
227	52	Poor
228	52	Poor
229	52	Poor
230	52	Poor
231	52	Poor
232	52	Poor
233	52	Poor
234	52	Poor
235	52	Poor
236	52	Poor
237	52	Poor
238	52	Poor
239	52	Poor
240	52	Poor
241	52	Poor
242	52	Poor
243	52	Poor
244	52	Poor
245	52	Poor
246	52	Poor
247	52	Poor
248	52	Poor
249	52	Poor
250	52	Poor
251	52	Poor
252	52	Poor
253	52	Poor
254	52	Poor
255	52	Poor
256	52	Poor
257	52	Poor
258	52	Poor
259	52	Poor
260	52	Poor
261	52	Poor
262	52	Poor
263	52	Poor
264	52	Poor
265	52	Poor
266	52	Poor
267	52	Poor
268	52	Poor
269	52	Poor
270	52	Poor
271	52	Poor
272	52	Poor
273	52	Poor
274	52	Poor
275	52	Poor
276	52	Poor
277	52	Poor
278	52	Poor
279	52	Poor
280	52	Poor
281	52	Poor
282	52	Poor
283	52	Poor
284	52	Poor
285	52	Poor
286	52	Poor
287	52	Poor
288	52	Poor
289	52	Poor
290	52	Poor
291	52	Poor
292	52	Poor
293	52	Poor
294	52	Poor
295	52	Poor
296	52	Poor
297	52	Poor
298	52	Poor
299	52	Poor
300	52	Poor

Fig:3.3 After Pre-processing the Dataset

3.3 CALCULATION OF AQI Range

```
In [62]: def AQI_Range(x):
        if x<=50:
            return "Good"
        elif x>50 and x<=100:
            return "Moderate"
        elif x>100 and x<=200:
            return "Poon"
        elif x>200 and x<=300:
            return "Unhealthy"
        elif x>300 and x<=400:
            return "Very unhealthy"
        elif x>400:
            return "Hazardous"

df['AQI_Range'] = df['AQI'] .apply(AQI_Range)
df.head()
```

Fig:3.4 Calculation of AQI range

3.4 PREDICTION OF AIR QUALITY:

we spitted the data set into two parts of first 75% and rest 25% data into test and train datasets to identify the huge seasonal variations and trend. By using KNN classification algorithm, we can predict the air quality.

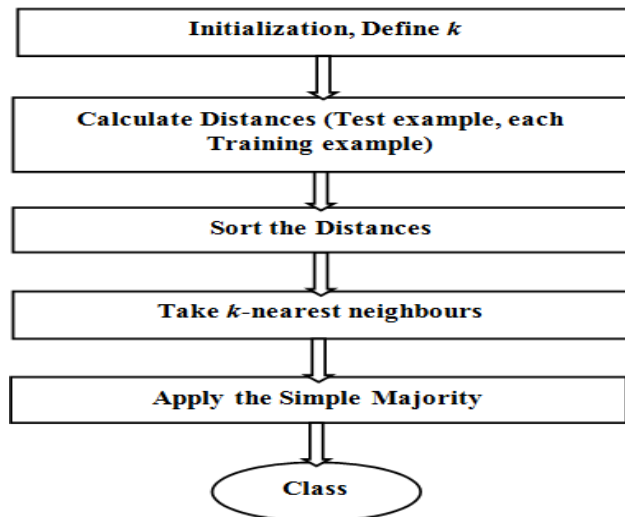


Fig:3.5 KNN Algorithm steps

3.5 RANDOM FOREST ALGORITHM:

Random Forest is a popular machine learning algorithm that belongs to the supervised learning technique. It can be used for both Classification and Regression problems in ML. It is based on the concept of ensemble learning, which is a process of combining multiple classifiers to solve a complex problem and to improve the performance of the model.

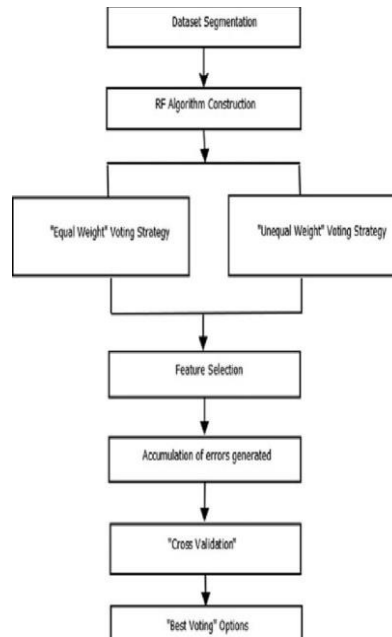


Fig:3.6 Random Forest Regression Algorithm

4.SYSTEM DESIGN

4.1 USE-CASE DIAGRAM:

A use case describes a function that a system performs to achieve the user’s goal. A use case must yield an observable result that is of value to the user of the system.

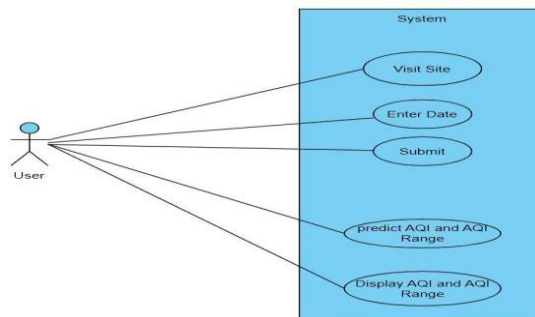


Fig:4.1 Use-Case Diagramm

4.2 CLASS DIAGRAM:

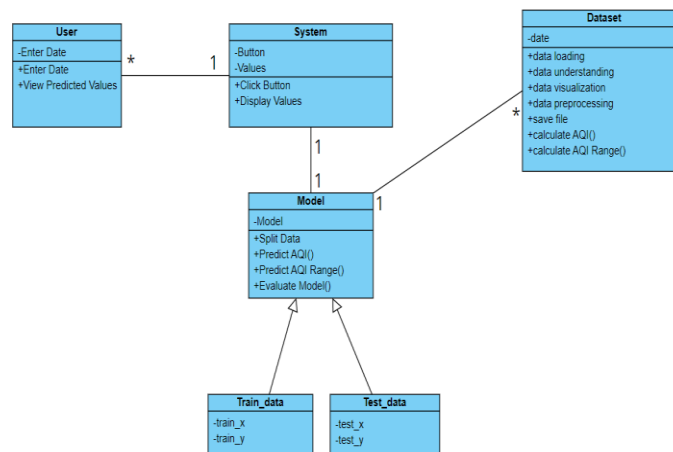


Fig:4.2 Class Diagram

4.3 ACTIVITY DIAGRAM:

Activity diagrams provide a way to model the workflow of a business process. An activity diagram is typically used for modeling the sequence of workflows

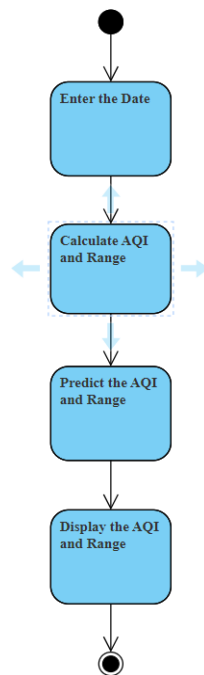


Fig:4.3 Activity Diagram

4.4 SEQUENCE DIAGRAM:

A Sequence diagram is a graphical view of a scenario that shows object interaction in a time- based sequence what happens first, what happens next. Sequence diagrams establish the roles of objects and help provide essential information to determine class responsibilities and interfaces.

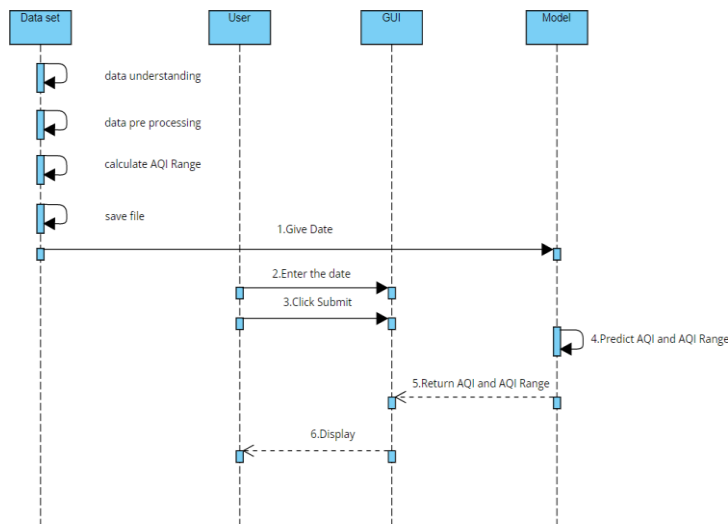


Fig:4.4 Sequence Diagram

5. METHODOLOGY

5.1 KNN ALGORITHM

Object model describes the structure of the system in terms of objects, attributes, associations, and operations. During requirements and analysis the object model starts as the analysis object model and describes the application concepts relevant to the system. K-Nearest Neighbour is one of the simplest Machine Learning algorithms based on Supervised Learning technique.

K-NN algorithm assumes the similarity between the new case/data and available cases and put the new case into the category that is most similar to the available categories.

K-NN algorithm stores all the available data and classifies a new data point based on the similarity. This means when new data appears then it can be easily classified into a well suite category by using K- NN algorithm.

K-NN algorithm can be used for Regression as well as for Classification but mostly it is used for the Classification problems.

K-NN is a non-parametric algorithm, which means it does not make any assumption on underlying data.

It is also called a lazy learner algorithm because it does not learn from the training set immediately instead it stores the dataset and at the time of classification, it performs an action on the dataset.

KNN algorithm at the training phase just stores the dataset and when it gets new data, then it classifies that data into a category that is much similar to the new data.

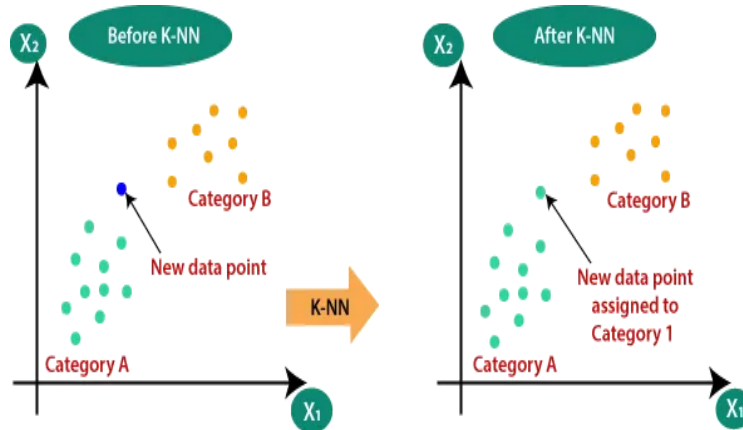


Fig:5.1 KNN Algorithm

5.2 RANDOM FOREST REGRESSION ALGORITHM:

Random Forest is a popular machine learning algorithm that belongs to the supervised learning technique. It can be used for both Classification and Regression problems in ML. It is based on the concept of ensemble learning, which is a process of combining multiple classifiers to solve a complex problem and to improve the model.

Random Forest works in two-phase first is to create the random forest by combining N decision tree, and second is to make predictions for each tree created in the first phase.

The Working process can be explained in the below steps and diagram:

Step-1: Select random K data points from the training set.

Step-2: Build the decision trees associated with the selected data points (Subsets).

Step-3: Choose the number N for decision trees that you want to build.

Step-4: Repeat Step 1 & 2.

Step-5: For new data points, find the predictions of each decision tree, and assign the new data points to the category that wins the majority votes.

Random Forest is capable of performing both Classification and Regression tasks. It is capable of handling large datasets with high dimensionality.

It enhances the accuracy of the model and prevents the overfitting issue

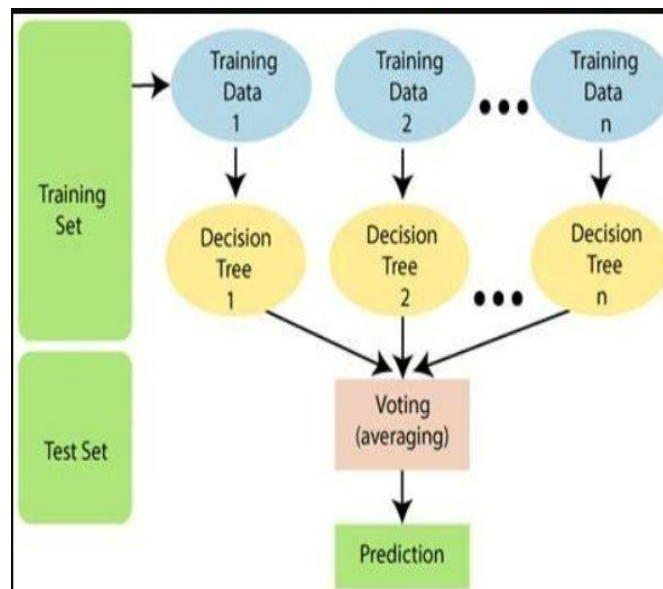


Fig:5.2 Random Forest Algorithm

5.3 CLASSIFICATION ALGORITHM

The Classification algorithm is a Supervised Learning technique that is used to identify the category of new observations on the basis of training data.

In Classification, a program learns from the given data set or observations and then classifies new observation into a number of classes or groups.

Such as, Yes or No, 0 or 1, Spam or Not Spam, cat or dog, etc. Classes can be called as targets/labels or categories.

Unlike regression, the output variable of Classification is a category, not a value, such as "Green or Blue", "fruit or animal", etc.

Since the Classification algorithm is a Supervised learning technique, hence it takes labeled input data, which means it contains input with the corresponding output.

The main goal of the Classification algorithm is to identify the category of a given data set, and these algorithms are mainly used to predict the output for the categorical data

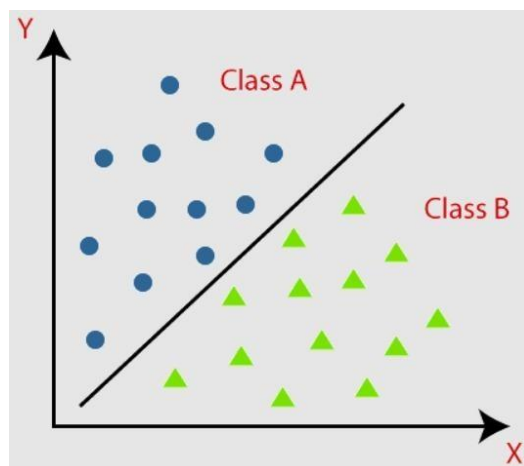


Fig:5.3 Classification Algorithm

6.OUTPUT SCREEN

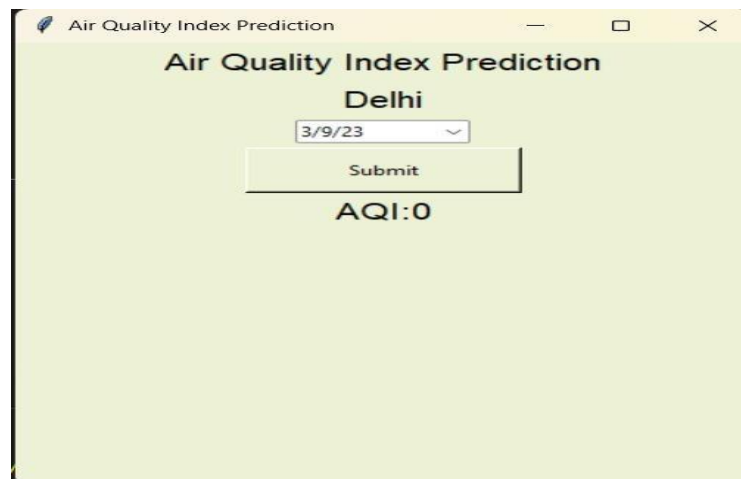


Fig:6.1 Before Giving the Date

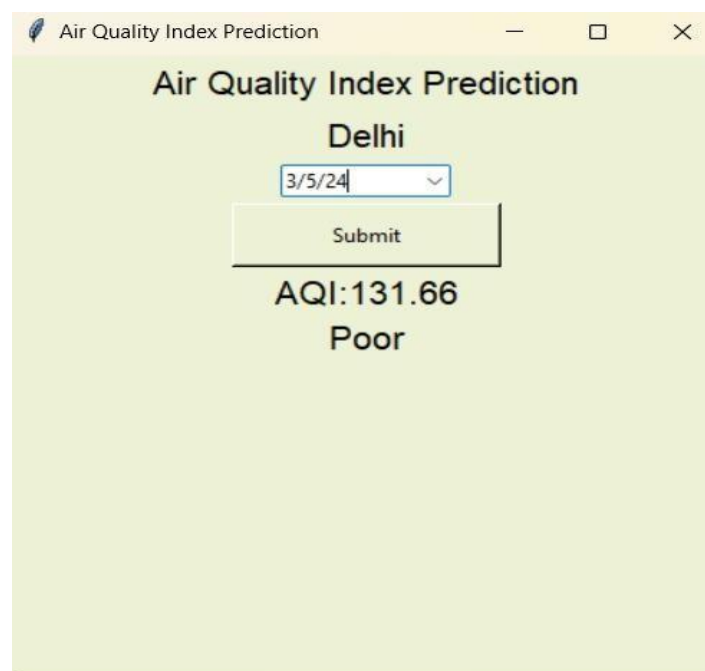


Fig:6.2 After Giving the Date

REFERENCES:

1. T M. Chiwewe and J. Ditsela, "Machine learning based estimation of Ozone using spatio- temporal data from air quality monitoring stations," 2016 IEEE 14th International Conference on Industrial Informatics (INDIN), 2016, pp. 58-63, doi: 10.1109/INDIN.2016.7819134.[1]
2. Masih A. Machine learning algorithms in air quality modeling. Global Journal of Environmental Science and Management. 2019 Oct 1;5(4):515-34.[2]
3. Amuthadevi C, Vijayan DS, Ramachandran V. Development of air quality monitoring (AQM) models using different machine learning approaches. Journal of Ambient Intelligence and Humanized Computing. 2021 Jan 3:1-3.[3]
4. Naidu DS, Aruna R. Study of Air Quality Detection using Machine Learning Techniques. International Journal of Scientific and Academic Research (IJSAR), eISSN: 2583-0279. 2022 Sep 16;2(8):1-8.[4]
5. S. Yarragunta, M. A. Nabi, J. P and R. S, "Prediction of Air Pollutants Using Supervised Machine Learning," 2021 5th International Conference on Intelligent Computing and Control Systems (ICICCS), 2021, pp. 1633-1640, doi: 10.1109/ICICCS51141.2021.9432078.[5]
6. Dragomir EG. Air quality index prediction using K-nearest neighbor technique. Bulletin of PG University of Ploiesti, Series Mathematics, Informatics, Physics, LXII. 2010 Jun 1;1(2010):103- 8.[6]
7. Mahalingam U, Elangovan K, Dobhal H, Valliappa C, Shrestha S, Kedam G. A machine learning model for air quality prediction for smart cities. In2019 International conference on wireless communications signal processing and networking (WiSPNET) 2019 Mar 21 (pp. 452-457). IEEE.[7]

8. Kingsy Grace R, Manju S. A comprehensive review of wireless sensor networks based air pollution monitoring systems. *Wireless Personal Communications*. 2019 Oct;108(4):2499- 515.[8]
9. Gallego E, Folch J, Teixidor P, Roca FJ, Perales JF. Outdoor air monitoring: Performance evaluation of a gas sensor to assess episodic nuisance/odorous events using active multi-sorbent bed tube sampling coupled to TD-GC/MS analysis. *Science of the total environment*. 2019 Dec 1;694:133752.[9]
10. Liu Y, Uthra DR. Bluetooth based smart home control and air monitoring system. *International Journal of Advanced Research in Engineering and Technology (IJARET)*. 2020;11(5).[10]
11. Goh CC, Kamarudin LM, Zakaria A, Nishizaki H, Ramli N, Mao X, Syed Zakaria SM, Kanagaraj E, Abdull Sukor AS, Elham MF. Real-time in-vehicle air quality monitoring system using machine learning prediction algorithm. *Sensors*. 2021 Jul 21;21(15):4956.[11].