

## Air Quality Prediction Using Classification Techniques

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### ABSTRACT

In several industry applications and metro areas nowadays, air internal control and restoration is one of the most significant practices. The quality of the air is harmed by various types of pollution caused by transportation, power, and petroleum usage, among other things. The evidence of noxious particles is posing significant consequence to the improving the overall quality of life in green infrastructure. It is important to integrate reliable environmental monitoring models that collect data on pollutant levels and provide an analysis of environmental pollution in each area compared to air pollution. Air pollution harms both the atmosphere and human health, resulting in global warming, acid rain, heart disease, lung cancer, and skin cancer among the population. More analysis is needed to assess raw information in view of identify pollutants. ML algorithms and deep learning ensure that potential AQI can be predicted and effective steps can be taken. As a result, air quality estimation and prediction has become a major research subject. We will look at various ML [26] and DL algorithms from which can identify air quality in this analysis.

### Keywords

Air quality index, Machine Learning, Deep Learning techniques, Air pollutants, Quality prediction.

### INTRODUCTION

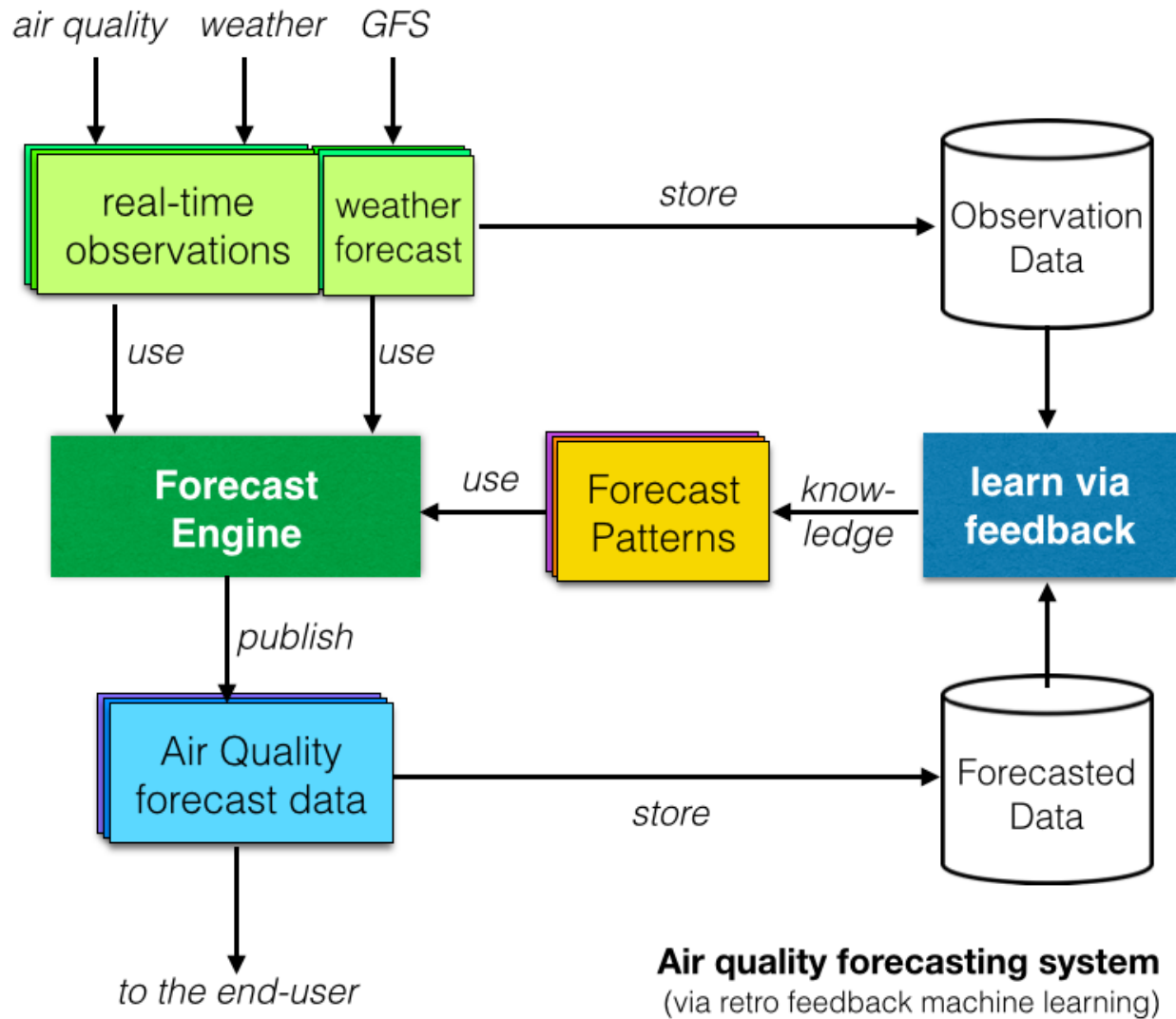
The statistics indicate that people are increasingly heading to towns. The United Nations (UN) reports, the Metropolitan People in 2018 would be around 55.3%, and it is expected to reach 68 percent by 2050. Urbanization brings with it a slew of issues relating to transportation, health care, air quality, and other facets of life. To address these issues, the smart city concept was created, which is a technology that combines information and communication (ICT) that residents as well as evolving assets to promote long-term growth and improved quality of life. There are many definitions to explain smart city, prefer 'A smart community is a metropolis where six key aspects include the intelligent economy, intelligent transportation, advanced environment, advanced citizenry, advanced life and procedural knowledge' or 'The use of sophisticated communication technology in order to make the smarter and more efficient city infrastructure and services such as city government, education, healthcare, real estate, public safety, transport and utilities.' Using intelligent city services, we can access a large number of stats to learn more about

the present situation to get a better view of the situation throughout the town. A key feature of urban mobility[25] is the accessibility of data provided by different sensors.

Air is amongst the most valuable ecological supporting the growth and survival of lifeforms. Air is essential for the survival of all types of life, including plants and animals. Consequently, every living organism needs high-quality, toxic-free air for survival. Because of the trend toward industrialization and a temporal long-term exposure interaction with fine mean a thing of particulates and acute rises through mortality, such as lung cancer and cardiopulmonary disease, air pollution has been the topic of many recent environmental studies. Throughout the years, traditional methods have been used to assess air quality. These approaches include manual data collection and analysis. Many researchers have suggested different solutions for predicting air pollution using mathematical or deep learning models, ranging from environmental to data science. At different times of the day, the parameters measured by meteorological stations (temperature, wind, humidity, and insolation) cloud form various set of potential features.

The aim of this paper is to review papers on estimation pollution of the air in digital machine learning will help cities algorithm, compare methodologies used by different authors, and get a general idea of applied approaches.

Deep learning techniques have begun to be actively established in this location or environment, and more surveys and assumptions were carried out conducted, that is dependent on the significance in the detention center. Combining all of it the data will aid in determining the trend and identifying the technologies used in the analysis survey setting, which then will guide us into the research consideration.



**Fig 1: Air quality prediction**

**RELATED WORK**

RUCHI RAUTRI, et.al.,[1] implemented the most interesting and toughest task is predicting air pollution making use of data mining. The majority of the systems in those days are intended to aid in the storage of air pollution data, analysis of that particular data and providing result in the form of statistics. Most of the systems make use of decision support systems, but their capabilities are severely limited. They will respond to basic questions such as "What is the highest range of the air pollution" and "Where are the most pollutants?" However, they are unable to answer complex queries such as "Predict subsequent month air pollutants count" or "Given me, tomorrow's air pollution data". This method employs prediction techniques.

BRAHIM K.K., et al., [2] developed a new framework of DL for analyze data from IoT[38,41] digital cities. We are proposing paper back method based on LSTM networks for

predicting hereafter air quality estimates in a digital city. The proposed model's result of the estimation is positive, indicating that it can be applied to other digital city complications in prediction as well.

PING-WEI SOH, et.al.,[3] goals to forecast air pleasant for as much as forty- eight the usage of a mixture of a couple of neural networks[28], which includes SNN, CNN, and LST period reminiscence in find temporal of spatial connections. Terrain extraction effect in air quality, proposed copy for prediction takes into account various meteorological data getting from the preceding couple of hours as well elevation space data. Observation from various locations are included in the model, which were derived from relationship between nearest areas and among comparable areas in domain of temporal. Observations with Taiwan and Beijing data units display suggested version higher overall outperforms contemporary trendy methods.

MENGLIRUAN, et.al.,[4] implemented an environment air reliability assessment and prediction copy focused in data mining. To assess environment air reliability, PCA is used to determine the principal component ratings. To design the environment air reliability model, data mining exponential smoothing method is used to process the input data. The results show that the designed method is comparable the real measurement range in surrounding quality of air and method has a better accuracy and better efficiency, which can be used as a benchmark for improving environmental quality.

XIUWENYI, et.al., [5] developed a DNN(Deep Neural Network) based method (entitled Deep air), it contains a spatial transformation aspect as well as a deep distributed network. Based on air pollutants spatial correlations, the previous part converts spatially sparse air quality prediction data into a coherent input that can re-create pollutant sources based on air pollutants spatial correlations. We deployed Deep Air in Air Pollution Prediction system, providing prediction of quality of air for more than three hundred cities of china in every hour. The result of the experiment is based on the 3-year and from 9-cities of china demonstrate benefit of Deep Air method which consists of ten base line. The percentage of short term[20], long term, sudden changes predictions are 2.4, 12.2, 63.2 respectively.

GAGANJOT KAUR KANG, et.al.,[6] investigated air quality forecasting making use of different big-data and machine learning (ML) based techniques. This paper summaries published research findings on prediction of air quality using AI, decision trees, ML, deep learning, and other technologies. Furthermore, it gives the solution to future research paper. It is suggested that it takes into account both spatial and temporal correlations. To extract inherent air quality features using a SAE (Stacked Auto-Encoder) method is used, which is trained greedily layer-wise.

YING ZHANG, et.al.,[7] resolved the issues of analyzing high-dimensional large-scale data. It is solved using LightGBM model. We could boost prediction accuracy by exploring the data forecasting[23,29,35,40] function and available data. we are compared more predicted[24] data after testing data in the dataset by comparing original data. The original data getting from Beijing which contains the 35 air quality monitoring stations. The sliding mechanism used to multi-dimensional secular property to increase the training data features in the dataset due to a lack of data.

DANIEL SCHURHOLZ, et.al.,[8] proposed a prediction model which uses context-aware data processing principles for combining an effective algorithm (Long Short-Term Memory and Deep Neural Network) which is used to predict the pollution of air and it gets data from pollution sources and the user's health profile. Outcomes acquired with My Air Quality Index -MyAQI show high accuracy levels are obtained (90% - 96%) and the usage of this model is very high to the user.

DITSUHIISKANDARYAN, et al.,[9] goal is to review articles on prediction of pollution level in air digital cities which is implemented by using ML algorithms, to compare methodologies used by different authors, a general idea about applied methods. Machine learning techniques have begun to be actively established in this area, several research and their result had conducted, and this is monitored based on the field's importance. The combination of all data will aid us in detecting trends and identifying developments in the experimental field, which will help us in our future experimental studies.

The main goal of this models is to review studies on prediction [17,22,27] of air pollution using ML algorithms which get the data from the sensor in the digital cities. Most of the important papers and articles were chosen using the most common databases.

ANTOINE ALLÉON, et al.,[10] concentrated on predicting the air pollution to given place and concentrated on air quality temporal variability. Air quality prediction is build by given monitoring station utilizing preceding measurements as well as weather forecasts. The past data are used to train the models, and the majority of recent papers that use this method use neural networks[30,37], specifically LSTM architectures. The four pollutants[11,15] expected by the engine are not always measured at each monitoring station. The experiments discussed in this paper were carried in Europe and the US, which are the two regions where we have obtained most accurate values.

S.NO	TITLE	TECHNIQUES	FINDINGS
1	Index Using Artificial Neural Network Recognition of Future Air Quality	Multilayer Perceptron	We present the prediction techniques that were used to provide next day and next month air pollution counts which is used to avoid problems. The prediction of air pollution using Data mining which is the most fascinating and challenging tasks, and we present the prediction techniques[18,19] that were used to provide next day and next month air pollution counts in order to avoid problems.

2	Air Quality Prediction-Deep Learning Model	LSTM[43]	The main goal of the model focused on LSTM- (Long Short Term Memory) network method for predicting air quality range in a smart city.
3	Air Quality Prediction Model (Most Relevant Spatial-Temporal Relations) using Adaptive Deep Learning (DL)	Convolutional Neural Network-(CNN)	The main goal is to forecast the quality of air for up to forty eight hours and a combination of various neural network[14,36] methods which includes ANN[12,31,33], LSTM, CNN [32]to implement seal-temporal connections.
4	Air Quality Evaluation and Prototype-Data Mining Techniques	Principal Component Analysis method	The use of data mining to verify and forecast ambient air quality is proposed. We may assess environmental air quality by calculating principal component ratings and using principal component analysis.
5	Air Quality Prediction-(Deep Distributed Fusion Network)	Deep Neural Network (DNN) algorithm	Proposed DNN method that includes a spatial transformation function it uses a Deep Distributed Fusion Network. A former portion takes into account the spatial associations of air pollutants
6	Air Quality Prediction-(Big Data & Machine Learning)	Genetic algorithm	It is important to introduce Successful air quality monitoring models that collect data on pollutant concentrations and provide an estimation of air pollution in each area are critical. As a result, air quality assessment and forecasting has grown in importance as a research subject.
7	Air Quality Prediction-Predictive Data Feature Exploration	Data fusion model	Use the LiRhtGBM model to solve the problem of analysing high-resolution of the large amount of data and as one of the data sources used to predict the air quality.
8	Air Quality Prediction-Artificial Intelligence	Long Short-Term Memory Deep Neural Network	Present a novel history prediction model that incorporates data from both surrounding pollution sources with an efficient air pollution prediction algorithm

9	Air Quality Prediction-Machine Learning (ML) Algorithms	Neural Network (NN)[44,45,46,47]	In the smart cities, our paper reviews studies on with the help of data measured by the sensor and with the algorithm of machine learning (ML) we were able to find the quality of air. The most important papers were chosen by making use of the common databases and also with filtration in the database.
10	Large-Scale Air Quality Forecasting-Convolutional LSTM Network	Convolutional LSTM	Given the close ties between weather and air pollution, develop air quality forecasts.

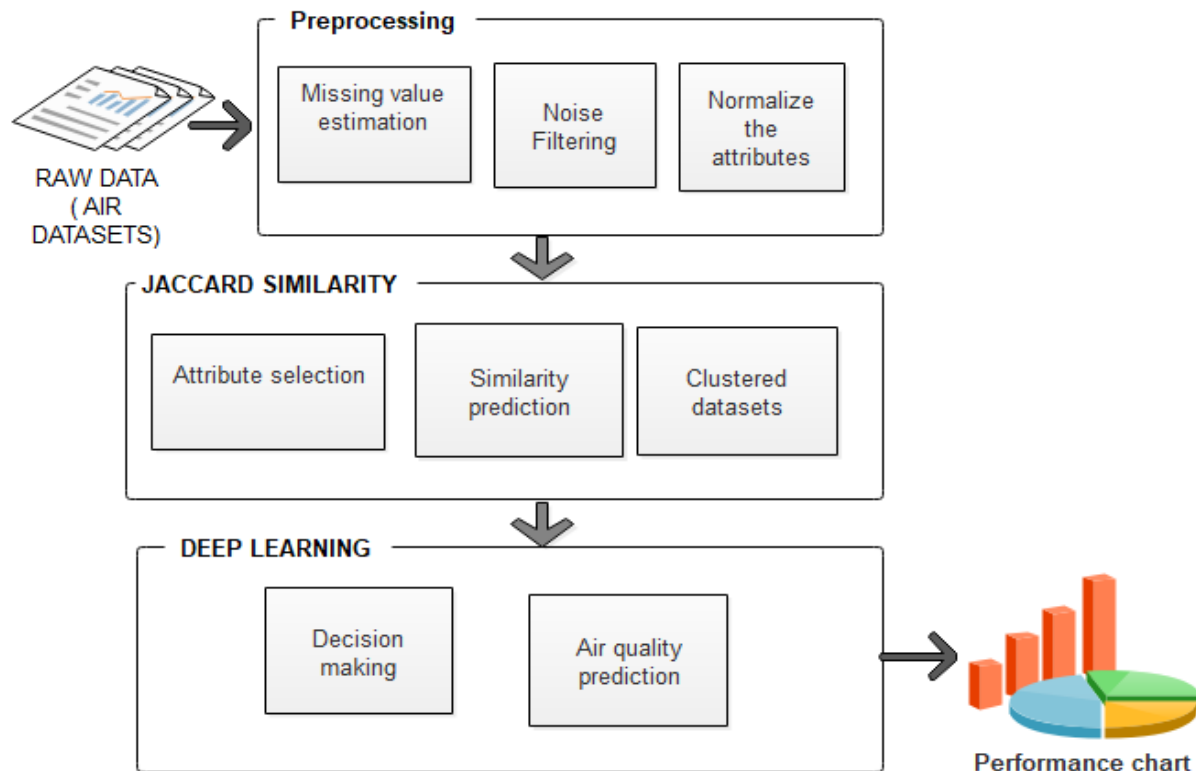
### PROPOSED FRAMEWORK USING DEEP LEARNING METHODS

Artificial Intelligence includes deep learning as a component. In machine learning (ML)[48], a program learns from data and knowledge using various computer algorithms. The computer does not need explicit programming. These will develop and modify the algorithm on their own. Governments in populated and developing countries view air control as a major challenge.

Air pollution is caused by traffic and weather conditions, the burning of fossil fuels, and industrial waste such as power plant emissions. Weather forecasting has traditionally been done by simulating the atmosphere as a fluid. It is taken a sample of the present state of the atmosphere. The potential state of the atmosphere is calculated by solving thermodynamic and fluid dynamics numerical equations. However, when calculating the initial conditions of the atmosphere, the conventional system of differential equations that governs the physical model can become unstable due to disturbances and uncertainties.

Deep learning is used to foresee the outcome based on historical evidence. Deep learning is an artificial intelligence division (AI) which allows computers to learn without needing to be specifically programmed. The attention of machine learning is on developing applications which can transform when exposed to new knowledge, as well as the fundamentals of machine learning[16], such as the implementation of a simple computer learning algorithm using the .NET system. The use of advanced algorithms is used in the coaching and prediction process. It feeds the training data to an algorithm, which then uses that experience to make predictions based on brand new test data.

The aim is to look into quality of air making use of deep learning. After pre-processing air quality[13,21,42] dataset, supervised deep learning is help to analyze the forecasting of air quality. Furthermore, deep learning methods for weather forecasting will be used. As a result, it is critical that we analyze this data in order to extract any valuable knowledge and purpose. Figure 2 depicts the basic structure of the diagram.



**Fig 2: Proposed framework**

## DATASETS ACQUISITION

A data set (or dataset, though many modern dictionaries, such as Merriam-Webster, do not use this spelling) is a broad data gathering. Determined from the data is typically the simple network table contents or statistical data statistics[34], where every table column presents a unique variable, and a specific datum member for each row. For each member of the datum, the datum lists values for all of the variables, such as deep and volume of the object. The data set may contain data through one or so more members depending on no's of members. The term datum may refers more loosely that indicates that the information contained in a set of related records that correspond to a specific analyze or event. We may upload air datasets with attributes including CO, Pb, SO<sub>2</sub>, PM, NO<sub>2</sub>, O<sub>3</sub>. We may upload weather datasets such as temperature, humidity, wind speed, and lighting values in this module. These datasets were gathered from the Irvine central location at University of California.

## PREPROCESSING

Data Preprocessing is the system of changing unprocessed raw information into a usable format. Some data is inaccurate, incomplete, contradictory and/or missing in certain trends. Data preprocessing is a tried and true method of addressing the existing challenges. Discarded Value changes are critical because if missing values aren't held as they should be, the data insights or overall performance of predictive models may be harmed. Imputation of lacking values from predictive techniques assumes that the person of such lacking observations is not decided at



random and that the variable selected to impute such lacking observations has a few courting with it, in any other case it is able to bring about vague estimates and remodel the dataset from unstructured layout to established layout. Then calculate the missing value evaluation and irrelevant data removal methods.

## **SIMILARITY MEASUREMENT**

Crossroads among both Union and the Jaccard coefficient of similarity make up the index of Jaccard. The Jaccard index equates two sets of values that would determine which are shared and which are unique. It would be the percentage compare of how similar the two different sets of data are, ranging from 0% to 100%. Greater the percentage, the more similar the two populations. Although it is simple to use, small sample dimensions are highly sensitive and can produce incorrect results, particularly with very small samples or datasets with incomplete comments. This module uses common measures to cluster datasets[37]. Grouping the traits or properties using JACCARD resemblance. The common metric between subnets is calculated using the Jaccard distance. The more dissimilar means bigger the “distance”. The greater the difference between the means, the more identical they are. Data is grouped and displayed on a page based on distances.

## **DEEP LEARNING PROCESS**

Deep Learning algorithms have the foreseeable future in the study of automated extraction of complex information descriptions in an automated manner. Deep Learning[39] computerized can learn and represent data in a layered and hierarchical manner. Through a hierarchical learning process, Deep Learning and Big Data Analytical techniques allows for the elevated retrieval, hybrid abstraction as data representation.

Big Data analysis is a significant advantage of Deep Learning because it can obtain insight from data and is a Big Data Analytics platform when large amount of manually curated information present. Based on attribute threshold checking, we can create the air quality index[15] using the CNN algorithm. We will create rules based on temperature, humidity, and wind speed in this module. The CNN algorithm is used to create the threshold values for each attribute. When the dataset is large and there are many variables to consider, the CNN algorithm is not as simple as it appears in our original example. Pruning is required at this point. Pruning refers to removing branches from our decision tree that do not contribute substantially to our decision-making process.

## **CONCLUSION**

Humans and plants also suffer from the impact of toxins in the air. Since there many different sources of emissions, determining the impact of air pollution on health is extremely difficult, and their individual effects vary greatly. The data has been preprocessed, and it can be further processed using a data mining tool to provide policymakers with appropriate decision support. This project demonstrates how data mining methods can be useful in the fields of environmental monitoring and natural resource research. We were used preprocessed data to find Air Quality Index. Data mining[50] tool has been used for modelling, forecasting, and decision-making support. Convolutional Neural Networks[32,49] were accustomed to analyze the data of the

Artificial Neural Network[12] model in Data Mining techniques. This study suggests that a deep learning algorithm can increase quality prediction performance.

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