

Smart Real-Time Indoor Air Quality Sensing System and Analytics

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Abstract - Indoor air quality monitoring and analytics is one of the important interdisciplinary research areas, which is attracting significant attention of various researchers from environment, mathematics, material science and electrical and computer engineering. According to a research study conducted by World Health Organization (WHO), pollution of indoor air is the most known hazardous case for respiratory diseases such as lung cancer, asthma and chronic diseases. Lack of information about the pollution sources and its serious impact on health leads to a huge number of people likely to be affected by various types of respiratory diseases. With the recent developments in sensing technology, machine learning and communication technology, IoT based Smart Real Time Indoor air quality sensing and analytics have been implemented to promote better awareness for users to alert them about indoor air quality to maintain the wellbeing in their indoor environments. The paper provides a proof of concept on IoT based Indoor air quality sensing system and analytics. The data is collected for analyzing indoor air quality in various indoor settings such as kitchen for oily based cooking, living room for insecticide spray, and smoking and flour mill for detecting flour dust during crop grinding. We used J48 and Naïve Bayes machine learning algorithm to model the air quality status. Result shows that the Naïve Bayes Algorithm detects 99.11% and J48 algorithm detects 99.30 % accurately

Keywords - Air Quality, Air pollution, Internet of Things (IoT), Indoor-air quality (IAQ), Smart Home, Smart sensing, machine learning algorithm.

I. INTRODUCTION

According to a research study conducted by WHO, indoor air pollution is the eighth most known risk factor for respiratory diseases such as lung cancer, asthma and chronic diseases. Lack of information about the pollution sources and its serious impact on health leads to a huge number of people likely to be affected by various types of respiratory diseases. With the recent developments in sensing technology, machine learning and communication technology, IoT based Smart Real Time Indoor air quality sensing and analytics have been implemented to promote better awareness for users to alert them about indoor air quality in order to maintain the wellbeing of the indoor inhabitants. Indoor air quality (IAQ) plays a vital role in our daily lives. Approximately 90 percent of people spend their time indoor environment. Study shows that poor indoor air quality is the leading cause of respiratory infections, chronic lung diseases, and cancers might cause significant risks to people's health as reported by World Health Organization (WHO). According to Environmental Protection Agency (EPA), indoor air pollution may be two to five times worse than the air outdoors. The major cases of indoor air pollution is daily household activities. For example, oil-based cooking could generate a harmful airborne particulate matter (PM), tobacco smoking produces various chemicals with many of them identified as poisons in the form of tiny particles (PM) and gases including volatile organic compounds (VOCs). Cleaning and maintenance products such as pesticides and disinfectants contain and releases PM 2.5 and many VOCs that lead to long- term health risks such as child development and hormonal issues [4], [5], [28], [30]. Even though we are highly exposed to the various type of indoor air pollution, we focused on the two reasons.

Major motivation of Indoor Air Quality monitoring is based on the fact that most of the indoor air pollutants are odorless and are not visible. Also, respiratory diseases have no simple symptoms until for long period of exposure [1]. Therefore, it is very important to determine and know whether the symptoms are due to exposure to indoor air pollutants or other related factors [2].

This paper presents an implementation of a IoT based system for the monitoring of indoor air pollution levels and their respective air quality indexes (AQI). In addition, the system could increase users' awareness towards IAQ, the information about the pollution detection, the AQI status or standard, the seriousness of the pollution, and the respective suggestions to help users how to reduce the pollution to its optimal level.

II. RELATED WORKS

Researchers have studied the problems related to air pollution using different technologies. Some of participatory sensing technique was developed for utilizing each person's mobile phone to create a collaborative sensor network to collect, share and analyze data.

In [7], [8] authors presented the concentration of carbon dioxide (CO₂) in indoor environment and analyzed the growth and decay rate of the pollutant. Brett J. et al. [9] shown that smoking is the most significant contributor to indoor air pollution in addition to a portion of the outdoor contribution.

Researchers also proposed a cloud-based approach comprising of particulate matter sensors and air quality analytics engine in the cloud for measuring PM_{2.5} in real time basis [10],[12]. In [11] Jong-J.etal. presented the role of sensor network for sensing various indoor air pollutants and airborne particles. They also discussed several factors that contributes to poor air quality in past and current building structure, so that the latter buildings are more airtight and hence brings in less fresh air from the outside.

Sunyoung et al. [12], also developed a homebased PM 2.5 monitoring system to visualize and share the real time indoor air information with others increasing people's awareness towards IAQ. **Haryono et al.** [13] emphasized on the pollutants emission from house hold activities and fuel based cooking especially liquid petroleum gas (LPG). They have shown that cooking ingredients and cooking methods strongly influences quality of indoor air. They have revealed that two typical cooking methods frying and boiling produces fine particles (PM_{2.5}) and CO.

Another researchers presented the rise of sensing technology for the energy management and IAQ in urban environments, also identified the major challenges for their large-scale deployment, and shown the research gaps that should be covered by future investigations along with the introduction of low-cost sensors technology [14]. In another study researchers also used the air quality data for building a model, based on artificial neural network to estimate how much hours an HVAC system should be turned on ahead of its original schedule to reduce indoor PM_{2.5} to a best situation [15].

Goncalo M. and Rui P. [16] developed an indoor air quality system based on an IoT paradigm for ambient assisted living to know a variety of environmental parameters such as air temperature, relative humidity, luminosity and concentration of carbon monoxide (CO), carbon dioxide (CO₂) gases.

Authors also developed air quality prediction model using neural network for environmental engineering problems [17] - [18]. Various parameters such as sulfur dioxide (SO₂), carbon monoxide (CO), nitrogen dioxide (NO₂), nitric oxide (NO), temperature, relative humidity and air velocity were used for modeling to minimize the increasing effects of air pollution. The concentration and composition of particles in indoor air are affected by both indoor and outdoor pollution sources.

Furthermore, in another machine learning approach, **Kingsy Grace. R et al.** in [21] presented analysis of air pollution using k-means clustering algorithm and compared the result with Possibility based Fuzzy C-Means clustering algorithm. The research work highlights that traditional methods are too complex to process and analyze the bulky data, and therefore heterogeneous data is converted into meaningful information by using data mining approaches for decision making.

Luis P. and Sanchez F. [24], presented models for assessment and prediction of air quality. Firstly, they developed an air quality assessment by using computational model for evaluating toxic compounds that affect sensitive people. They also proposed the model that predicts the concentrations of air quality by using an autoregressive technique to provide air quality index based on the fuzzy inference system previously developed. Another study in [25] found that smoking, candle, frying, grilling, stove use, toasting, cooking pizza, vaporizing oil and fan heater use could raise the indoor micrometer particle concentration levels.

Asmaa A. et al. also shown that the rise in concentration of CO₂ at a work place leads to a rise in the amount of volatile organic compounds (VOCs), hence decreases the performance of the workers in office [28].

III. METHODOLOGY

This section describes the methodology used in the study, the hardware and software components, threshold based algorithm, ML algorithm, and analysis of data generated by sensors, android app used for visualizing the sensor data. The air quality sensors, microcontroller and other important devices used for the study are explained in detail. **Architectural design:** The crucial components used in the system development are described in the following section.

A. Sensing components

The IAQ Sensing layer consists of the air quality sensors, those are PM 2.5 and VOCs, CO, Temperature and humidity sensors.

Table 1. Sensing platform

Air Quality Sensor	Standard name
Particulate Matter (PM 2.5)	Sharp GP2Y1010AU0F
Carbon monoxide (CO)	MQ-7
Volatile Organic Compounds (VOCs)	MQ-135
Humidity	DHT22
Temperature	DHT22

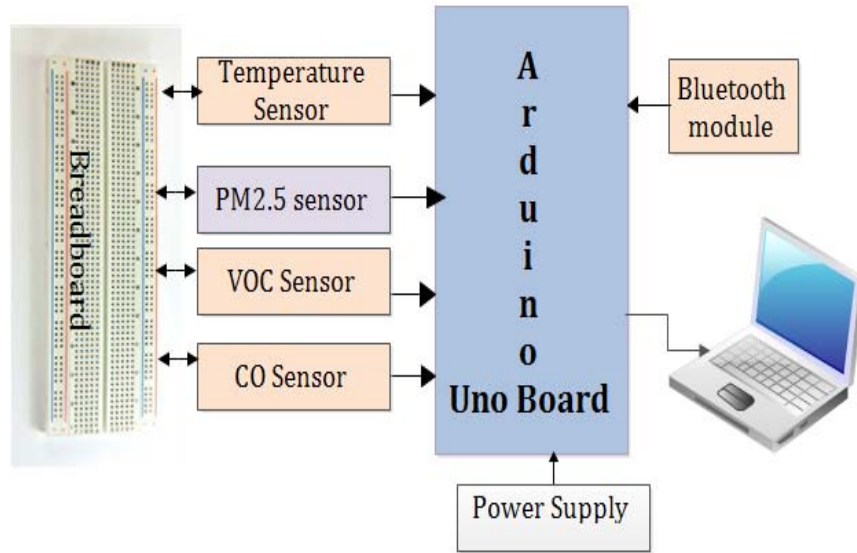


Fig.1 Block diagram for hardware interfacing.

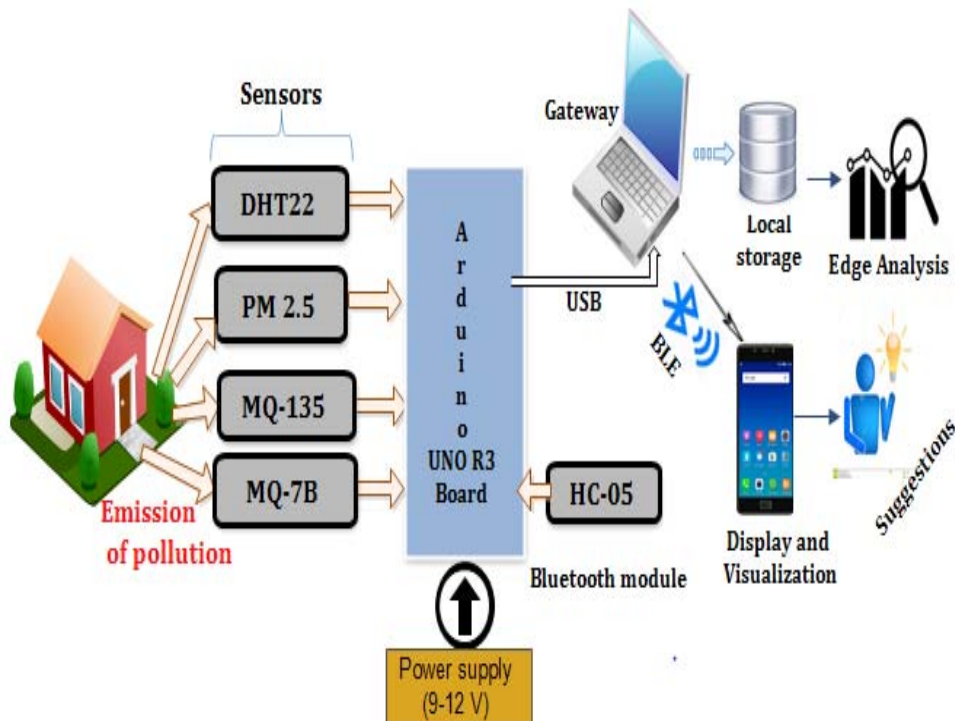


Fig.2 Proposed system architecture

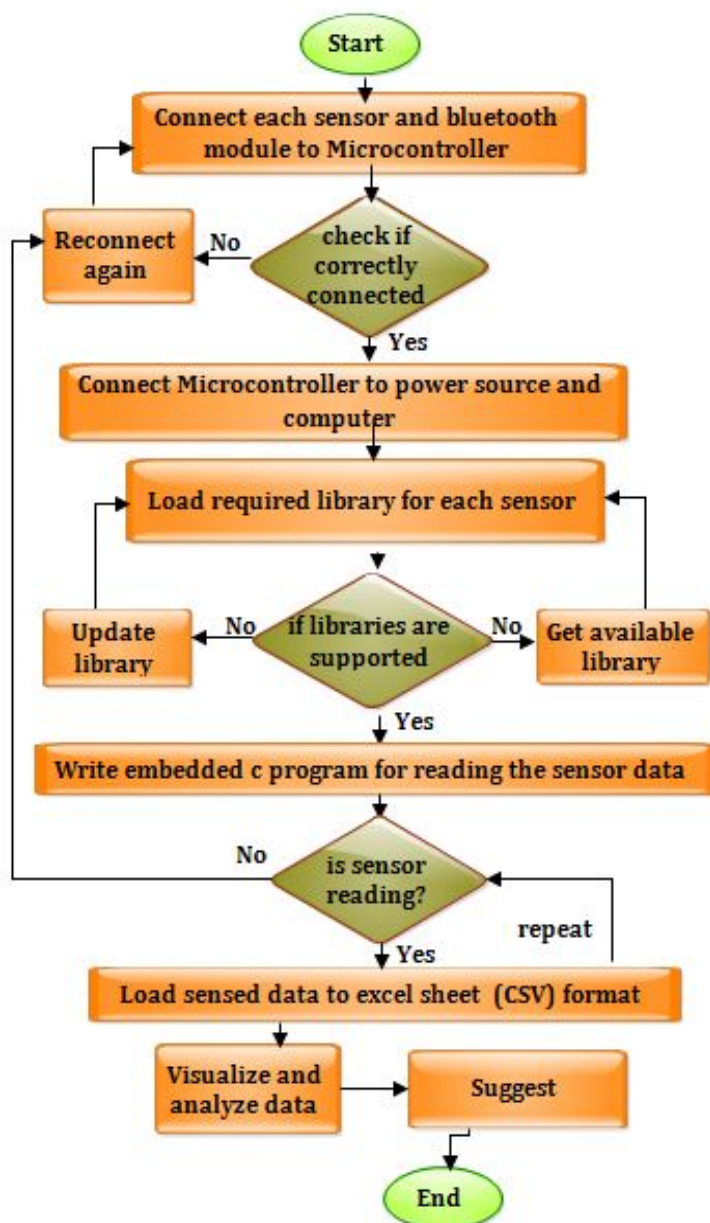


Fig.3 flow chart of the system

The air quality sensors are described in the following sections.

1) *MQ-135 Air quality sensor*: this measures general air quality because it is sensitive to many gases and VOCs (Volatile Organic Compounds) including formaldehyde, benzene, ammonia (NH₃), nitrogen oxides (NO_x), alcohol, smoke and carbon dioxide (CO₂). We used this sensor for sensing VOC pollutions.

2) *MQ-7 Carbon monoxide (CO) sensor*: used for detection of CO concentration. The detection Range is 0-500ppm (part per million). The Application area is domestic CO gas leakage alarm. The MQ-7 gas sensor is SnO₂, which with lower conductivity in clean air. It detects CO at low temperature (heated by 1.5V). The sensor's conductivity gets higher along with the CO gas concentration is rising. Figures shown below are the hardware components used for implementation of the Smart Real Time Indoor Air Sense

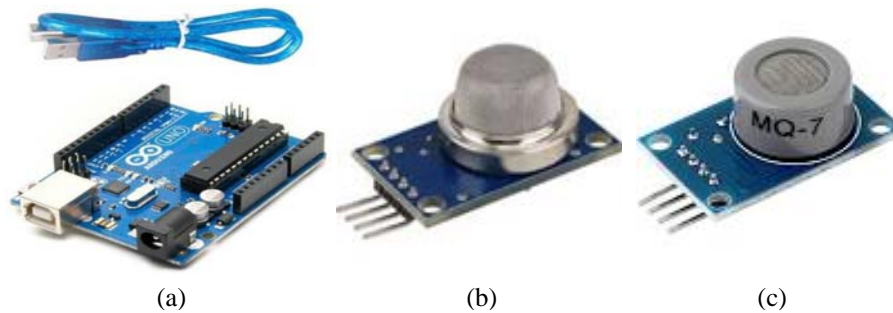


Fig. 4 (a) Arduino Uno R3 microcontroller board (b) MQ-135 sensor (c) MQ-7 CO sensor

3) *DHT22 Temperature and humidity sensor*: DHT22 is a digital, temperature and humidity sensor. It uses a capacitive humidity sensor and a thermistor to measure the surrounding air, and generate a digital signal on the data pin (no analog input pins used). It is shown in fig. 5(a)

- Cost effective having 4 pins
- Uses 3 to 5V power
- Good for -40 to 125°C temperature readings
- Good for 0-100% humidity readings

4) *Sharp GP2Y1010AU0F*: This is an optical air quality sensor, designed to sense dust particles with small size.

An infrared emitting diode and a phototransistor are diagonally arranged into this device, to allow it to detect the reflected light of dust in air. It is effective in detecting very fine particles like cigarette smoke, and also commonly used in air purifier systems. It is shown in fig. 5(b)

5) *HC-05 Bluetooth module*: It is IEEE 802.15.1 standardized protocol, through which one can build wireless Personal Area Network. HC-05 Bluetooth module is designed for wireless communication. It uses serial port to communicate with microcontroller devices. It is used for many applications like wireless headset, game controllers, wireless mouse, wireless keyboard etc. It has range up to <100m. It is shown in fig. 5(c)

B. Connecting and interfacing equipment's

1) *Breadboard*: is a device used as a construction base in developing an electronic circuit. It is a good unit for making temporary circuits and prototyping. It is made of plastic having strips of metal underneath with a numerous holes and it is solder less, which allows them to be reusable.

2) *Jumper wires* are an electrical wire with a connector or pin at each end. Used to interconnect the components of a breadboard with other equipment or components, without soldering.

3) *Resistors*: are used to reduce current flow, adjust signal levels, divide voltages, and terminate transmission lines.

4) *Capacitors* are used for storing energy electrostatically in an electric field. It also used for power conditioning. We used capacitor 220 μ F (microfarad) with PM2.5 sensor.

C. Processing and communication components

1) *Arduino Uno R3 microcontroller*: an arduino is a microcontroller used for building and interfacing various sensors and devices required for a given project. It allows uploading sketches/programs into the microcontroller memory. Arduino consists of both a physical programmable circuit board (often referred to as a microcontroller) and a software, or IDE. It is shown in fig. 4(a).

2) *Gateway device*: we used the personal computer as a gate way.

3) *Bluetooth technology*: It is IEEE 802.15.1 standardized protocol, through which one can build wireless Personal Area Network. HC-05 Bluetooth module is designed for wireless communication. It uses serial port to communicate with microcontroller devices. It is used for many applications like wireless headset, game controllers, wireless mouse, wireless keyboard etc. It has range up to <100m.

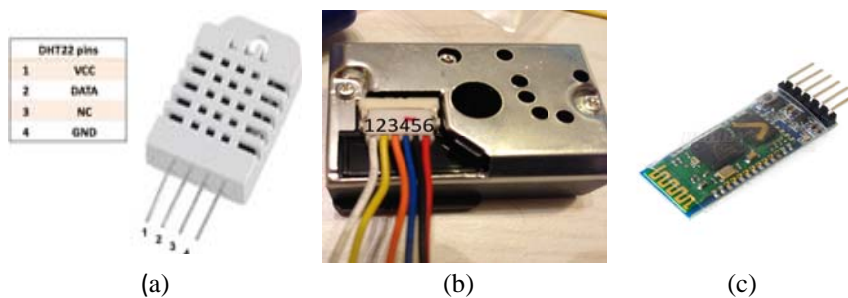


Fig. 5 shows the hardware components used for implementation of the SmartRealTimeIndoorAirSense (a) DHT22 temperature and Humidity sensor (b) Sharp GP2Y1010AU0F PM 2.5 sensor (c) HC-05 Bluetooth module

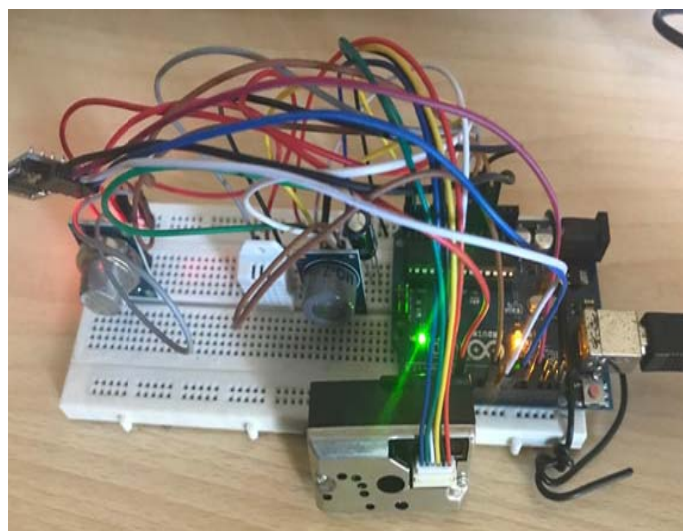


Fig 6 a proof of concept of SmartRealTimeIndoorAirSense

IV. ANALYSIS AND VISUALIZATION

The EPA (Environment Protection Agency) uses the Air Quality Index (AQI) to measure five major pollutants for which it has established National Ambient Air Quality Standards under the Clean Air Act. AQI is a numerical scale developed by EPA. It is used by government agencies to communicate how polluted or clear the air is in our surrounding for reporting day to day air quality with regard to human health. An increase in air quality index indicates increased air pollution and severe threats to human health [5], [40].

There are six distinct AQI categories: Good, Moderate, unhealthy for Sensitive Groups, Unhealthy, Very Unhealthy, and Hazardous. The threshold value for each category were developed based on established national air quality standards [32]. AQI calculations focus on major air pollutants including particulate matter, carbon monoxide (CO), sulfur dioxide (SO₂), ground-level ozone, and nitrogen dioxide (NO₂). The sub-index value is calculated for each of these pollutants. The highest sub-index reflects overall AQI. The general formula used to calculate air quality index [33] is given by equation (1).

$$AQI = ((I_{Hi} - I_{Lo}) / (BP_{Hi} - BP_{Lo})) * (C_p - BP_{Lo}) + I_{Lo} \dots \dots \dots (1)$$

Where

- AQI = the AQI for each pollutant
- I_{Hi} = the AQI value corresponding to BP_{Hi}
- I_{Lo} = the AQI value corresponding to BP_{Lo}
- C_p = the rounded concentration of the pollutant
- BP_{Hi} = the breakpoint that is greater or equal to C_p
- BP_{Lo} = the breakpoint that is less than or equal to C_p

A. Experimental Setup

Data Collection: the data collection has been completed in two weeks in three various environments to obtain the sufficient amount of data. We deployed our sensor module during home based activities especially cooking in the kitchen, smoking and spraying of insecticide in the living room and grinding flour mill. The sensor module was kept around two to five meter (2-5m) near to the cooking, smoking, spraying and grinding activities to detect the emission of pollution. The generated data was stored to comma delimited value (CSV) file format for further analysis. The features on the CSV file are temperature, humidity, PM2.5, VOC, CO, AQI and Air quality status.

B. ML algorithm for classification of the air quality

WEKA is a suite of machine learning software tool having various standard data mining tasks such as data preprocessing, regression, clustering, association rules, classification, prediction, feature selection and visualization. After gathering air quality data, we performed the preprocessing such filling or cleaning missing data values, removing inconsistent data points.

1) *Naive Bayes Machine Learning Algorithm*: This algorithm is a classification algorithm used for binary (two-class) and multi-class classification problems. The technique is easiest to understand when described using binary or categorical input values. We applied the naïve Bayes algorithms to our data set, the algorithm detects **99.11%** instances correctly as healthy and **0.89%** as unhealthy.

2) *J48 Machine Learning Algorithm*: This is one of best predictive and classification ML algorithm. We used this ML algorithm for accurately classifying the AQI category either healthy or unhealthy depending on the available features of the data set. We followed the EPA AQI standard to group the AQI categories of good and moderate as healthy and group the remaining four AQI categories as unhealthy [32]. The total data set contains 3043 instances. The algorithm detects **99.31%** instances correctly as healthy and **0.69%** as unhealthy.

Result of Naïve Bayes Algorithm

Time taken to build model: 0.44 seconds

=== Stratified cross-validation ===

=== Summary ===

Correctly Classified Instances	3022	99.3099 %
Incorrectly Classified Instances	21	0.6901 %
Kappa statistic	0.861	
Mean absolute error	0.0109	
Root mean squared error	0.0813	
Relative absolute error	20.0995 %	
Root relative squared error	49.6363 %	
Total Number of Instances	3043	

=== Detailed Accuracy By Class ===

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	0.999	0.202	0.994	0.999	0.996	0.864	0.887	0.993	healthy
	0.798	0.001	0.944	0.798	0.865	0.864	0.887	0.818	unhealthy
Weighted Avg.	0.993	0.197	0.993	0.993	0.993	0.864	0.887	0.988	

=== Confusion Matrix ===

a	b	<-- classified as
2955	4	a = healthy
17	67	b = unhealthy

Results of J48 Algorithm

Time taken to build model: 0.09 seconds

=== Stratified cross-validation ===
 === Summary ===

Correctly Classified Instances	3016	99.1127 %
Incorrectly Classified Instances	27	0.8873 %
Kappa statistic	0.8375	
Mean absolute error	0.0088	
Root mean squared error	0.0939	
Relative absolute error	16.3829 %	
Root relative squared error	57.3409 %	
Total Number of Instances	3043	

=== Detailed Accuracy By Class ===

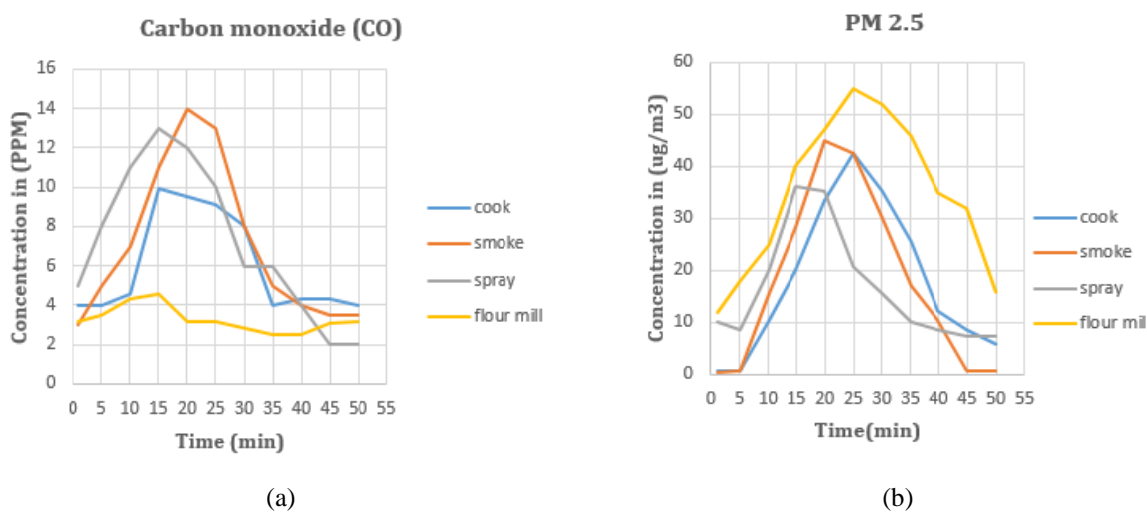
	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	0.995	0.143	0.996	0.995	0.995	0.838	0.918	0.996	healthy
	0.857	0.005	0.828	0.857	0.842	0.838	0.917	0.769	unhealthy
Weighted Avg.	0.991	0.139	0.991	0.991	0.991	0.838	0.918	0.989	

=== Confusion Matrix ===

```

a   b  <-- classified as
2944 15 | a = healthy
  12 72 | b = unhealthy
    
```

The pollution event due to various indoor activities cooking, smoking, and insecticide spray are visualized by the graphs in fig. 7



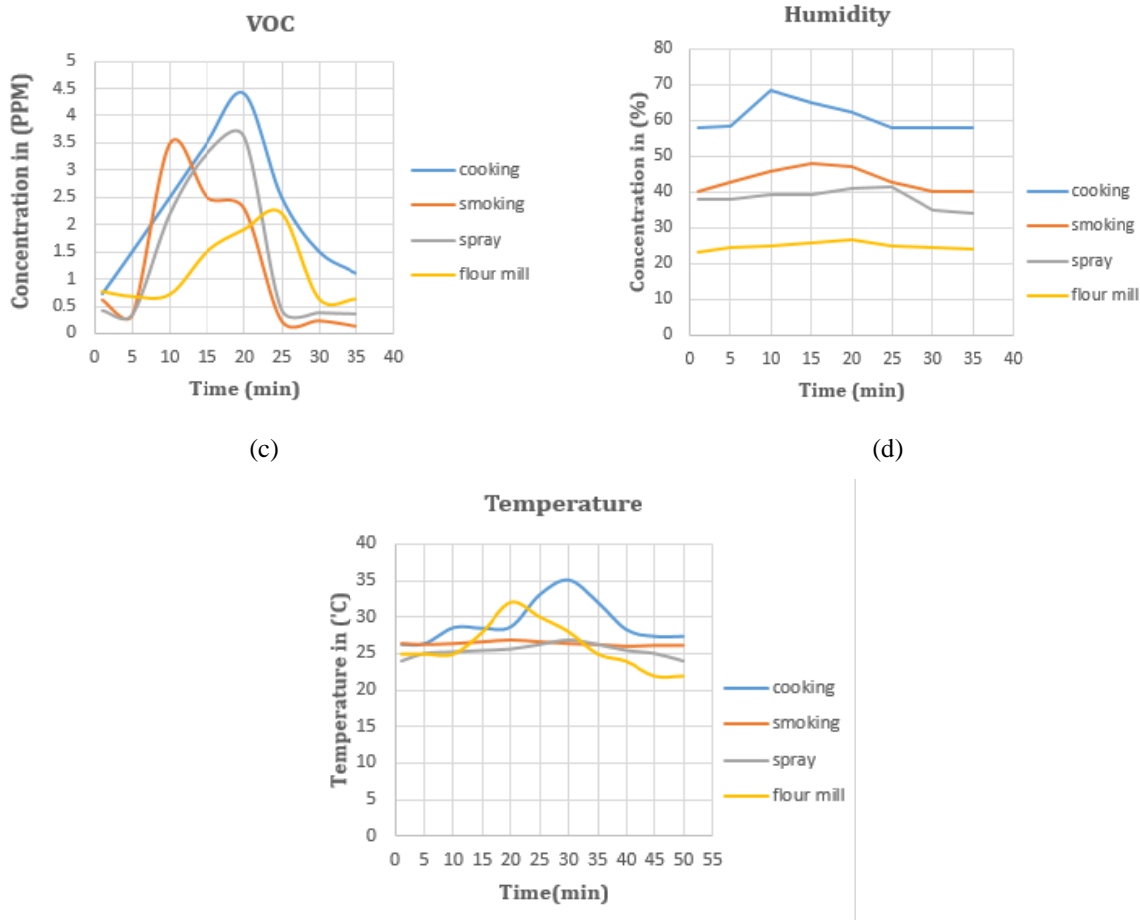


Fig. 7. The above graph shows (a) CO, (b) PM 2.5, (c)VOCs, (d) humidity sensor and (e) Temperature measurements of indoor air pollution generated by cooking, smoking, insecticide spray and flour dust.

C. Smart phone application

The goal of developing an android application is for increasing user awareness of IAQ, visualizing the lively streaming indoor air quality data within a specified interval of time. After installing and launching the application, the login window will be displayed. The user must fill the required credentials to login to the application. If the user wants to exit, it is possible to use the Cancel button. After logging in the main layout will be displayed. Then establishing the Bluetooth connection, hence the data from the sensor will be transferred to smart phone through serial port by Bluetooth communication protocol. Depending on the amount of the pollution, AQI and each pollutant concentration, suggestion to users will be displayed accordingly.

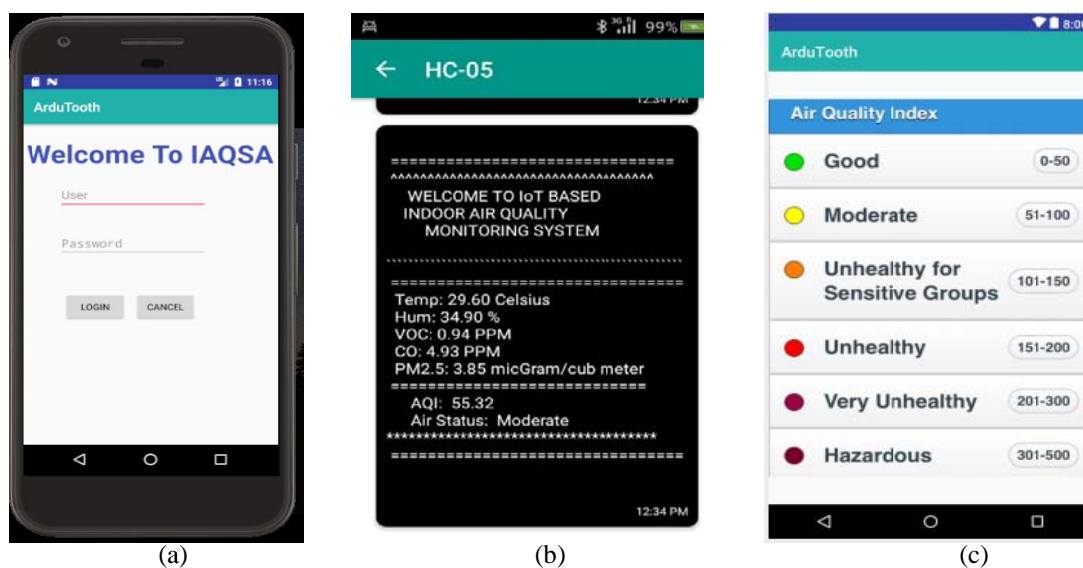


Figure 8. The SmartRealTimeIndoorAirSense an android applicationscreenshot of (a) login window, (b)visualizing the lively streaming indoor air quality information and suggestion (c) the EPA based AQI standard with color code.

V. CONCLUSION

In this work, we designed a prototype of SmartRealTimeIndoorAirSense, an intelligent real time indoor air quality sensing, monitoring and analytics system that detects indoor air pollution events, visualizing pollution levels of each pollutant to estimate the AQI, and providing actionable suggestions to help user in order to improve IAQ in a timely manner. Specifically, SmartRealTimeIndoorAirSense is developed using cost effective air quality sensors for continuously monitoring the concentrations of CO, PM 2.5 and VOCs, which are three of the most common air pollutants in the indoor environment and the ambient temperature and humidity sensors for sensing temperature and humidity level at home. SmartRealTimeIndoorAirSense is currently designed for indoor home uses, but it can be extended to be used in public buildings such as office rooms, class rooms, shopping malls and subway stations. Therefore, we believe SmartRealTimeIndoorAirSense has incredible chance to be adopted widely in a real world.

The study further highlights the strong link between indoor air quality and possible health risk in the form of various respiratory diseases. Despite suitable measures and guidelines defined by agencies including WHO, we believe that the air pollution is the serious ongoing issue which requires further attention.

VI. FUTURE DIRECTIONS

We suggest that the responsible authority and Environment protection agency should work with researchers, Information Technology companies, government's authorities, to minimize air pollution and its severe causes. The research work on indoor air quality can be extended in several directions:

With the ongoing focus on Smart City challenges including air pollution in India, **further research work is required to improve participatory sensing** techniques used to create a map of air quality information.

Future research is needed to develop cost-effective sensors capable of measuring low concentrations of gaseous pollutants and distribution of nanoparticles. This could be of interest to various researchers and industry professionals active in nanotechnology and nanoparticles.

The problem of finding a correlation between the air quality and possible health risks in terms of respiratory diseases, has been studied widely. But there is no single correlation measure which is better than the other. Also, Indoor Air Quality is relatively new as compared to outdoor air pollution.

Therefore, **further research can be conducted to develop new correlation measures** to improve the correlation of indoor air pollutants with diseases. This could attract the attention of a mathematician working with environment researcher.

For professionals in computer science and IT domain, an IoT platform can be developed using fog computing and cloud computing to do the device level analytics and analytics at the air pollution control room respectively.

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