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## Proceeding Paper Climate Monitoring and Black Carbon Detection Using Raspberry Pi with Machine Learning <sup>+</sup>

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Abstract: The Proposed climate monitoring system aims to address the substantial risks to human 10 health, climate stability, and ecological balance posed by air pollution, utilizes Raspberry Pi as a 11 central procession unit and integrates various sensors, which also incorporate sensors to measure 12 the concentrations of PM1, PM2.5, PM10, and black carbon. This method meets the need for effective 13 and immediate air quality monitoring and offers useful information to communities, academics, 14 and policymakers. Through IoT connectivity, the gathered data is sent to a cloud-based platform for 15 analysis and visualization. The system offers a user-friendly interface that presents actionable in-16 sights for informed decision-making. Its warning capabilities alert users when pollution levels ex-17 ceed thresholds and also this system contributes to a comprehensive understanding of air pollution. 18 By measuring particulate matter and black carbon levels, it supports the development of effective 19 air quality management strategies. The system helps to take proactive measures and create cleaner 20 and healthier environments. In conclusion, the proposed Climate Monitoring System utilizing Rasp-21 berry Pi, sensors, IoT connectivity, and machine learning techniques offers an effective and real-22 time solution for monitoring air quality. The integration of IoT connectivity allows remote access to 23 air quality data, while machine learning algorithms analyse the data and initiate alerts. 24

Keywords: Climate; Raspberry pi; sensors; IOT

## 1. Introduction

In this paper, we explore machine learning algorithms in the context of air pollution 28 (AP) prediction and monitoring utilizing data from Internet of Things (IoT) sensors. While 29 machine learning algorithms have rapidly advanced and found applications across vari-30 ous fields and domains, the realm of AP prediction remains open for exploration. One 31 particularly crucial parameter, particulate matter, serves as a pivotal indicator of pollution 32 levels in specific areas at precise times. These essential pollutant data are gathered 33 through sensor networks. Particulate matter as a vital parameter, offers a distinct and un-34 equivocal signal of pollution levels in the area at that specific time. These pollutant data 35 are obtained through sensor-based extraction. 36

To process the data from these sensors, an analog-to-digital converter is necessary 37 since most of them produce analog output. The Raspberry Pi 3B+ microcontroller is em-38 ployed to handle this task using specialized software and coding. The data is then ana-39 lyzed, and a graphical representation illustrating changes in the local environment over 40 time is generated. The validity of the results was confirmed through verification. The ex-41 periment control authority. This system has the potential to facilitate real-time decision-42 making and is particularly effective in addressing the ongoing issue of elevated air pollu-43 tion levels in many Indian cities [1,2] 44

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**Copyright:** © 2023 by the authors. Submitted for possible open access publication under the terms and conditions of the Creative Commons Attribution (CC BY) license (https://creativecommons.org/license s/by/4.0/). In this project, we suggest a technique to address this problem using a Raspberry Pi 46 board and a few sensors. MQ2, MQ135, and DHT11 can measure air temperature, humidity, and gas concentration. The BMP180 sensor will measure air pressure. These sensor 48 measurements are continuously transferred to the cloud platform (shown in **Figure 1**). 49 Consequently, using this technology makes it feasible to continuously monitor the meteorological conditions. Data will be shown on the LCD. We require an ADC to send data 51 from the gas sensors to the Raspberry Pi because the sensors generate analog values. 52

Web applications in embedded systems offer remote device and system monitoring and control using a web-based interface The integration of machine learning algorithms empowers the system to analyze the collected data [3,4], detect patterns, and identify potential air pollution sources or trends. With its user-friendly interface and actionable insights, this advanced air quality monitoring system using Raspberry Pi serves as a valuable tool for individuals, communities, and organizations in safeguarding and improving air quality for a healthier and sustainable future [5,6].

## 3. Principle of Operation

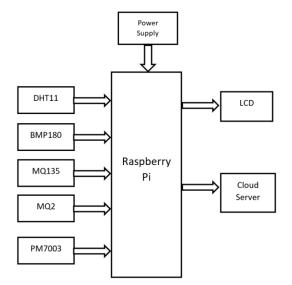
2. Data and Methodology

Machine learning in embedded systems is the integration of machine learning algo-61 rithms and techniques into small, resource-constrained devices like microcontrollers, IoT 62 devices, or embedded systems. This makes it possible for these devices to carry out intel-63 ligent functions, make predictions, and modify their behaviour in response to inputs with-64 out needing constant online access. These devices can handle and analyse data in real-65 time by integrating machine learning capabilities locally, making them more responsive 66 and effective. Robotics, autonomous vehicles, smart home gadgets, industrial automation, 67 and healthcare are just a few examples of the many fields where machine learning in em-68 bedded systems finds applications[7,8]. To achieve effective and precise performance 69 within the confines of the device, machine learning models must be deployed on embed-70 ded systems, which necessitates careful consideration of CPU resources, power consump-71 tion, memory constraints, and model optimization strategies[9]. Web applications in em-72 bedded systems offer remote device and system monitoring and control using a web-73 based interface The integration of machine learning algorithms empowers the system to 74 analyze the collected data, detect patterns, and identify potential air pollution sources or 75 trends[10,11]. With its user-friendly interface and actionable insights, this advanced cli-76 mate monitoring system using Raspberry Pi serves as a valuable tool for individuals, com-77 munities, and organizations in safeguarding and improving the weather around us for a 78 healthier and sustainable future[12,13]. 79

| 3.1. Components Used   | 80 |
|--|----|
| Raspberry Pi   | 81 |
| MQ2 Sensor   | 82 |
| DHT11 Sensor   | 83 |
| • LCD  | 84 |
| BMP135 Gas Sensor Module   | 85 |
| PM7003 Sensor  | 86 |
| 4. Results   | 87 |
| The results of the Climate Monitoring of Block Carbon Detection using Raspberry Pi | 88 |
| with Machine Learning are shown in these results Figures 2, 3, and 4.              | 89 |

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Figure 1. illustrates the Block diagram of our Climate Monitoring system using machine learning91and Raspberry Pi.92



Figure 2. Photograph of the LCD display showing the real-time temperature and humidity read-94ings.95



Figure 3. Photograph of the LCD screen showing the CO and CO2 values.





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## 5. Conclusion

This project proposes the implementation of an IOT system utilizing a Raspberry Pi 101 microcontroller for monitoring and improving the air quality of the environment. The uti-102 lization of IOT technology enhances the monitoring of various environmental factors, in-103 cluding the air quality issue the air quality issue addressed in this project. Here, the use 104 of temperature, and gas sensors and black carbon particles provides the feel of various 105 harmful gases that regulate the entire operation. In order to solve the problems with air 106 quality, the integrated IoT air pollution system was created. Various hazardous gases that 107 are present in the environment are mostly detected by the sensors. Additionally, these 108 sensors can be implemented in mobile automatons that can detect the earth's polluting 109 gases. By introducing new conventions, security measures can be improved to secure the 110 ata being delivered across the segments 111

|   | data being delivered across the segments.  | 111               |
|---|--|-------------------|
|   | Supplementary Materials: Not applicable  | 112               |
|   | <b>Author Contributions:</b> M.C. developed the entire hardware using Python code with the scientific guidance of M.V.L. carried out the data analysis and wrote the manuscript. Every author has reviewed and consented to the final published version of the manuscript. | 113<br>114<br>115 |
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|   | Data Availability Statement: In This study work we are using real-time sensor data.  | 119               |
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|   | Conflicts of Interest: The authors have no conflict of interest to disclose.   | 122               |
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