

Designing Machine Learning Software for Fragrance-Based Air Quality Optimization

Abbas Abdulazeez Abdulhameed¹, Yasmin Makki Mohialden², Nadia Mahmood Hussien³, Ethar Abdul Wahhab Hachim^{*4}, Salwa Hameed Naser Al-Rubae⁵

^{1,2,3,4} Department of Computer Science, College of Science, Mustansiriyah University, Baghdad, Iraq

⁵ Department of Chemistry, College of Science, Mustansiriyah University, Baghdad, Iraq

E-mail: abbasabdulazeez@uomustansiriyah.edu.iq, ymmiraq2009@uomustansiriyah.edu.iq,

nadia.cs89@uomustansiriyah.edu.iq, ethar201124@uomustansiriyah.edu.iq, drsalwahnaser@uomustansiriyah.edu.iq

Keywords: air quality optimization, machine learning, fragrance diffusion

Received: April 4, 2025

This paper presents the design of a machine learning-based software system for fragrance-based indoor air quality optimization, integrating advanced sensors and adaptive algorithms to enhance environmental conditions and user comfort. The proposed framework is a conceptual model that utilizes real-time data acquisition, predictive modeling, and intelligent fragrance diffusion for dynamically regulates air quality and humidity. By leveraging machine learning techniques, the system analyzes air pollutants, personalizes fragrance profiles, and optimizes indoor environments for energy-efficient air purification. Unlike conventional air quality monitoring systems, this approach actively enhances indoor air quality rather than merely detecting pollutants. The system's adaptive fragrance diffusion mechanism ensures cost-effectiveness and broad applicability across various indoor settings. However, as this study focuses on system design without practical implementation, further validation through prototype development and empirical evaluation is essential to assess its feasibility and effectiveness. Future research should explore real-world testing, additional AI-driven optimization techniques, and integration with IoT-based smart environments to refine the proposed system's capabilities.

Povzetek: Članek predstavlja zasnovo programskega sistema na osnovi strojenega učenja za optimizacijo kakovosti zraka v zaprtih prostorih z uporabo dišav. Sistem dinamično uravnava zrak in vlažnost s senzorsko akvizicijo podatkov, napovednim modeliranjem in prilagodljivim razprševanjem dišav za udobje uporabnikov.

1 Introduction

Up to 90% of urban residents spend most of their time indoors, increasing their risk of developing Sick Building Syndrome (SBS), a condition caused by poor indoor air quality [1]. Indoor air pollution ranks among the top five environmental threats to public health, significantly affecting respiratory, cognitive, and overall well-being [2]. Ensuring optimal humidity levels and clean air is crucial for individuals at home and in workplaces.

Traditional indoor air quality (IAQ) monitoring systems primarily detect environmental parameters, such as temperature, humidity, pressure, particulate matter (PM), carbon monoxide (CO), carbon dioxide (CO₂), oxygen (O₂), ozone (O₃), and volatile organic compounds (VOCs). However, these systems focus solely on detection rather than actively improving air quality.

This research proposes the design of a machine learning-based software system for fragrance-based indoor air quality optimization. The system is conceptualized to integrate sensor networks, predictive machine learning algorithms, and adaptive fragrance diffusion mechanisms. Unlike traditional IAQ monitoring systems that passively observe air quality, this software-driven approach is intended to dynamically

adjust air conditions and enhance user comfort through intelligent fragrance management. As this study focuses on design rather than implementation, future research should validate the practical feasibility and real-world effectiveness of the proposed system through prototype development and empirical testing.

2 Research contributions

This research presents a conceptual framework for a sustainable, software-driven air quality management system incorporating machine learning. The key contributions of this study include:

- a. **Software-Driven Fragrance Monitoring:** A novel software architecture that integrates machine learning and sensor networks for real-time IAQ assessment with adaptive fragrance diffusion.
- b. **Non-Intrusive Air Quality Optimization:** Unlike bulky filtration-based solutions, the proposed software system leverages intelligent fragrance diffusion to enhance indoor air quality.
- c. **Human-Centered Algorithm Design:** The system personalizes fragrance diffusion using adaptive AI models, improving user comfort and air quality based on preferences.

d. Machine Learning Integration: The software employs predictive analytics and adaptive learning algorithms to optimize fragrance release based on real-time sensor data.

e. Scalability and Future Applicability: The conceptual design is intended for homes, offices, and healthcare facilities, with potential for IoT integration and smart environment adaptation.

Traditional IAQ monitoring systems are often costly and disruptive, requiring frequent maintenance and high-maintenance filtration mechanisms. This study introduces a software-based intelligent air quality management approach, leveraging machine learning-driven fragrance diffusion to improve air conditions efficiently. The system design relies on:

- Python programming for data integration and sensor interfacing.
- Raspberry Pi as a potential embedded computing unit for future prototyping.
- Machine learning libraries (NumPy, Pandas, Matplotlib) for data analysis and decision-making.

Environmental monitoring via sensor-based machine learning models is a rapidly evolving field with applications in healthcare, smart homes, and industrial settings. The proposed software-based fragrance air quality management system represents a scalable and sustainable IAQ solution with future potential for real-world testing.

The remainder of the paper is structured as follows:

Section 2 Reviews related works on air quality monitoring and fragrance-based environmental management, Section 3 Presents the conceptual design of the machine learning-based software system for air quality optimization. Section 4 discusses the potential effectiveness and scalability of the proposed system, Section 5 Concludes with findings, limitations, and future research directions.

3 Literature review

Indoor air quality (IAQ) influences health and comfort as city folks spend most time indoors. New sensor tech and AI offer ways to monitor and improve IAQ. This review examines the role of fragrance sensors, machine learning, and integrated systems in managing indoor spaces.

3.1 Fragrance sensors in indoor air quality monitoring

The functioning principle of fragrance sensors mimics human olfaction to identify VOCs and pollutants thereby making evaluations of indoor air quality possible. The air quality monitoring system from Integra Fragrances uses an immediate detection method to track environmental changes for creating pleasant and clean indoor air environments. A new system will merge sensory devices with a dynamic fragrance distribution framework that controls environmental comfort by responding to present environmental patterns [17].

3.2 Machine learning in indoor air quality management

The implementation of IAQ management systems with integrated computer algorithms analyzes sensor data to enhance forecasting abilities and enhance indoor control activities [17]. The proposed system uses real-time data about environmental conditions and user ratings for learning how to modify its fragrance delivery. Building ventilation experiences improvements because AI systems help optimize HVAC systems while real-time IAQ monitoring provides safety benefits and decreased energy consumption.

3.3 Integrated air quality monitoring solutions

Health information becomes accessible through advanced IAQ monitors such as Airknight 9-in-1 together with Airthings 4200 Kit, which combine multiple sensors to track pollutants and allergens and smoke contamination and other contaminants for supporting sustainable comfort solutions [18,19]. The proposed system would utilize combined monitoring solutions to operate the adaptive fragrance release mechanism, which automatically adjusts fragrance output based on air quality and user preferences for improved indoor comfort and air quality.

4 Proposed method

The proposed Fragrance-Based Environmental Monitoring System (FBEMS) is designed to enhance indoor air quality and humidity by integrating various technologies that monitor and adjust environmental conditions. Table 1 represents the detailed description of the system's components:

Table 1: System components

Component	Description
Fragrance Sensors	Advanced sensors detect and analyze air quality indicators like PM2.5, PM10, VOCs, CO, CO₂, and various odors in various indoor environments. They also monitor humidity to provide a comprehensive environmental assessment.
Humidity Sensors	- These sensors measure atmospheric moisture levels to maintain an optimal indoor environment. - They ensure proper humidity levels, which are crucial for comfort and air quality.
Data Processing Unit (DPU)	- Acts as the central processing hub, receiving real-time data from fragrance and humidity sensors. - Utilizes machine learning algorithms to analyze environmental conditions and determine appropriate corrective actions.
User Preferences	- The system allows users to customize fragrance intensity, duration, and air quality settings. - User preferences are integrated into

	the data processing algorithm to personalize the experience.
Adaptive Learning Module	• Employs machine learning techniques to improve system performance over time. • Analyzes historical data to enhance predictions and optimize fragrance release strategies.
Fragrance Release Mechanism	• Based on data analysis and user preferences, the system releases customized fragrances at appropriate intensities and durations. • Ensures a pleasant indoor ambiance while maintaining optimal air quality.

This method aims to combine multiple sensors and adaptive systems to improve indoor air quality by continuously monitoring and adjusting the environment based on real-time data.

Figure 1 illustrates the Fragrance-Based Environmental Monitoring System, displaying the interaction between sensors, data processing, and fragrance release mechanisms. The system operates as follows:

- **Fragrance and humidity sensors** collect environmental data in real-time, monitoring factors like air quality and humidity levels.
- The **Data Processing Unit** analyzes the data, taking into account air quality, humidity, and user preferences to determine optimal conditions.
- The **Adaptive Learning Module** improves the system's efficiency over time by learning from historical data and refining its predictions.
- Based on the processed analysis, the **Fragrance Release Mechanism** disperses fragrances at the appropriate intensity and duration.
- Users receive **real-time updates** and can easily adjust system settings via an intuitive interface, available on both mobile app and website platforms.

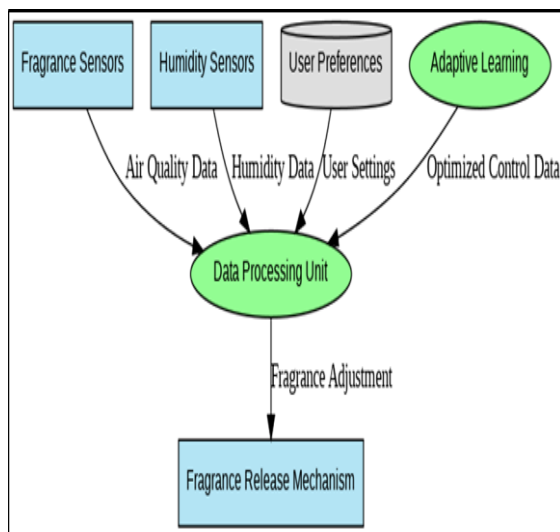


Figure 1: Proposed fragrance-based environmental monitoring system

While figure 2 shows the fragrance-based environmental monitoring system flowchart. It demonstrates a linear approach from data collection to constant monitoring and modifications.

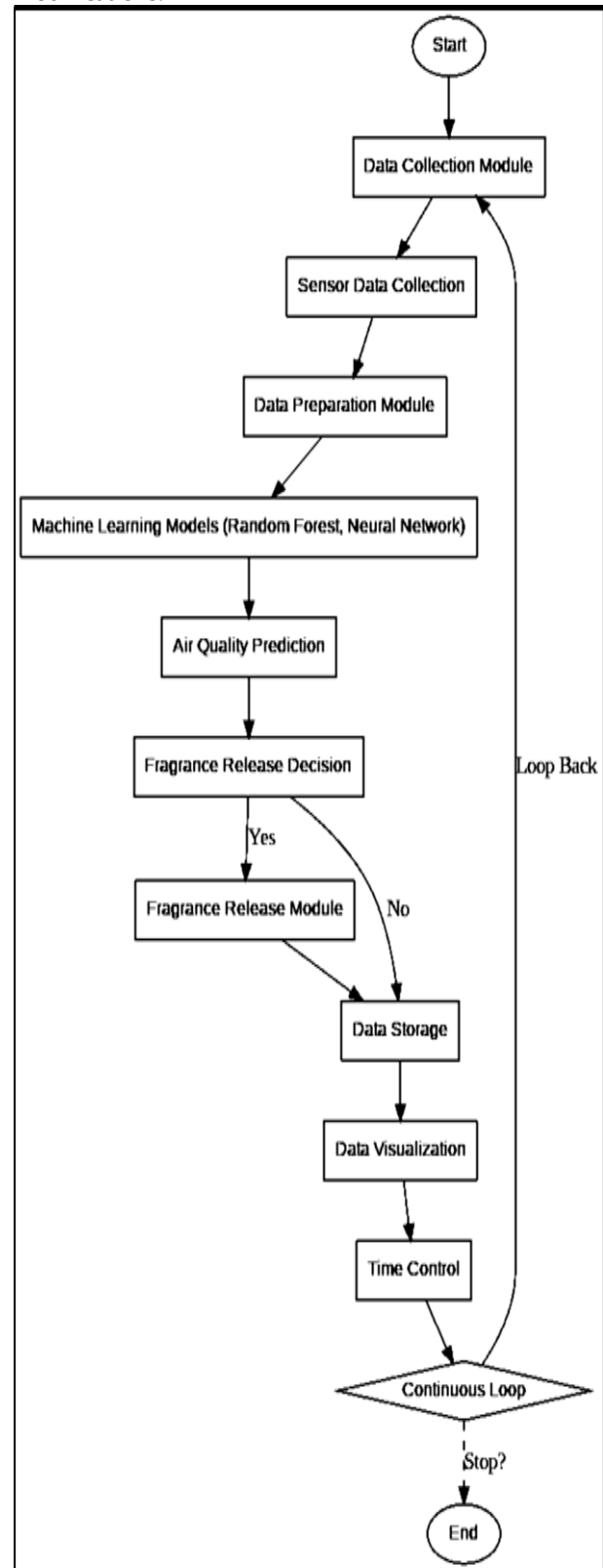


Figure 2: Proposed fragrance-based environmental monitoring system flowchart.

The flowchart shows a continuous cycle of data gathering, preparation, analysis, decision-making, and action. This method monitors and optimizes indoor air quality using sensor data and machine learning predictions in real time.

Table 2 show A Detailed Description of Each Component and Their Interactions:

Table 2: proposed system components and their interaction

Component	Description
Data Collection Module	Gathers data from various environmental sensors to start the procedure. Measures humidity, temperature, and air quality.
Sensor Data Collection	Captures raw data from fragrance-detection sensors, humidity sensors, and air quality monitors, including temperature, humidity levels, and airborne contaminants.
Data Preparation Module	Filters, preprocesses, and structures sensor data for further analysis, ensuring accuracy and compatibility with machine learning models.
Machine Learning Models	Evaluates prepared data using advanced machine learning techniques. Trains and optimizes Random Forest and Neural Network models for accurate air quality forecasting.
Air Quality Prediction	Predicts air quality in real time using trained machine learning models. Assesses contaminant levels and other environmental factors.
Fragrance Release Decision	Analyzes air quality forecast to determine if fragrance release is necessary. Considers user-defined preferences and threshold values.
Fragrance Release Module	Activates scent diffusion mechanisms when needed. Adjusts type, intensity, and duration of fragrance release to optimize indoor air quality.
Data Storage	Logs fragrance release actions, environmental conditions, and system responses in a database or CSV file. Enables performance review and enhancements.
Data Visualization	Transforms data into interactive graphs, dashboards, and real-time visual reports for pattern recognition and trend analysis.
Time Control	Manages scheduling of data collection, processing, and system updates to ensure uninterrupted monitoring and adjustments.
Continuous Loop	Operates continuously, analyzing environmental changes and making real-time adjustments to maintain optimal air quality.

A diagram in Figure 3 Display Explanation of Component Interactions it demonstrates the operations of the fragrance-based environmental monitoring system together with its fundamental features and connectors. Under the Main Function, the system operates through sensors for data acquisition and releases fragrance following data storage in the system.

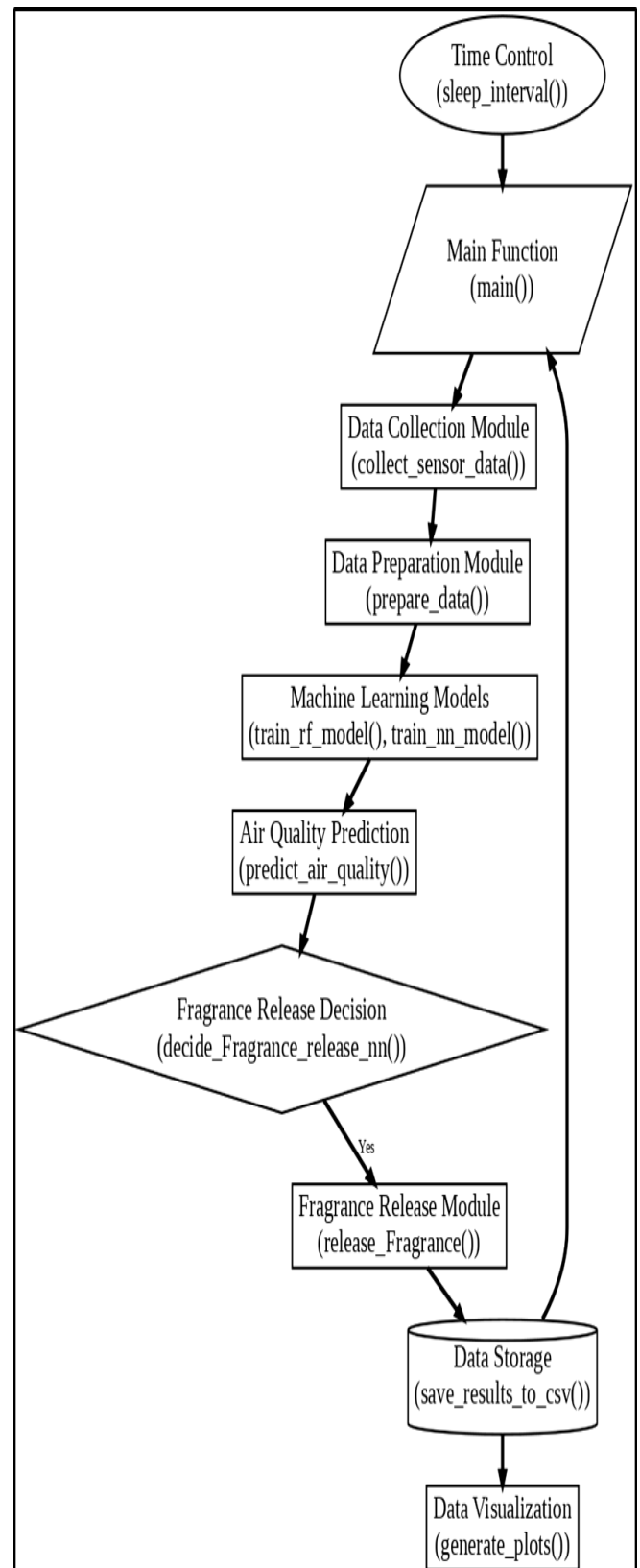


Figure 3: Operations of the proposed system.

The module and function chart presents the system's data processing and movement sequence. Figure 4 shows the interaction between modules in the proposed system.

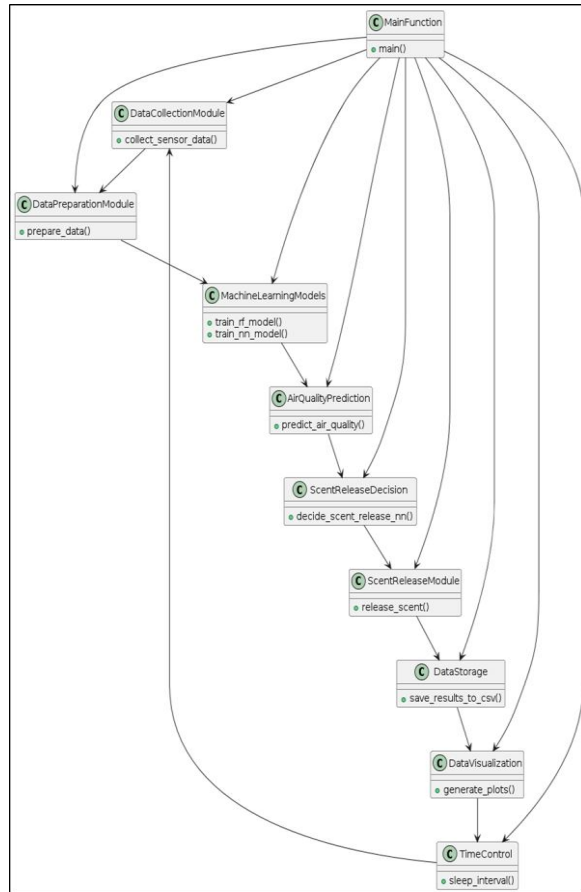


Figure 4: The interaction between modules in the proposed system.

4.1 Fragrance-based air quality assessment

The main operational aspect of the fragrance-based environmental monitoring system involves conducting instant air quality evaluations. The sensor network tracks indoor air quality through regularly placed fragrance sensors, which detect pollution, and changes in humidity in individual rooms. The system operates as follows:

- a. Through real-time operations, the fragrance sensors monitor airborne pollutants by tracking particulate matter levels of PM_{2.5} and PM₁₀ together with volatile organic compounds (VOCs) and carbon monoxide (CO) as well as carbon dioxide (CO₂) and detect various odors. The system records humidity values with environmental monitoring data to deliver extensive building conditions.
- b. Advanced machine-learning programs inside the Data Processing Unit study information received from the data collection stage. The system employs pre-established thresholds for classification, which reveals potential risks and establishes what actions need to be taken in response.
- c. The system responds by activating fragrance diffusion following any detection of subpar air quality measurements that go beyond acceptable standards to diminish pollutants while improving air freshness levels.

4.2 Humidity control

Optimal air moisture leads to both environmental comfort and indoor health levels. The fragrance-based environmental monitoring system applies smart humidity control to several sequential operations.

a. The system consists of humidity sensors, which execute continuous measurement of air moisture levels until the Data Processing Unit receives current humidity status information.

b. Cost-effective analysis of humidity data with air quality measurements helps the system decide proper moisture level adjustments.

c. The system enables fragrance-based humidity control because it automatically activates fragrance emission when comfort zones become affected by humidity changes.

The system maintains balanced environment quality through its control of scent intensity together with dispersion speed, which leads to proper indoor humidity restoration. The system displays its commitment to indoor air quality sustainability through its assessment of environmental quality along with humidity control measures.

4.3 The proposed system algorithm

The fragrance-based environmental monitoring system operates through a structured algorithm consisting of the following key steps as shown in Table 3.

Table 3: Algorithm steps

step	Description
Step 1: Initialization	Activate sensors, set fragrance diffusion parameters, and establish sensor communication.
Step 2: Data Collection	Continuously gather data on air quality, temperature, and humidity from sensors and external APIs.
Step 3: Air Quality Assessment	Analyze collected data to determine air quality based on AQI calculations.
Step 4: Humidity Control	Monitor and regulate humidity levels, triggering fragrance diffusion if necessary.
Step 5: Fragrance Release Decision	Determine optimal fragrance release based on air quality, humidity, and user preferences.
Step 6: Fragrance Diffusion	Activate and adjust fragrance diffusion to improve air quality.
Step 7: User Interaction	Allow users to provide feedback and override system decisions when needed.
Step 8: Monitoring and Feedback	Track indoor environmental changes and alert users to any critical issues.
Step 9: System Optimization	Continuously improve performance using machine learning models and historical insights.

5 Use case scenarios

Figure 5 illustrates a series of use-case scenarios that highlight the adaptability and versatility of the Fragrance-Based Environmental Monitoring System. These scenarios demonstrate how the system can be applied in various indoor environments to improve air quality and overall comfort.

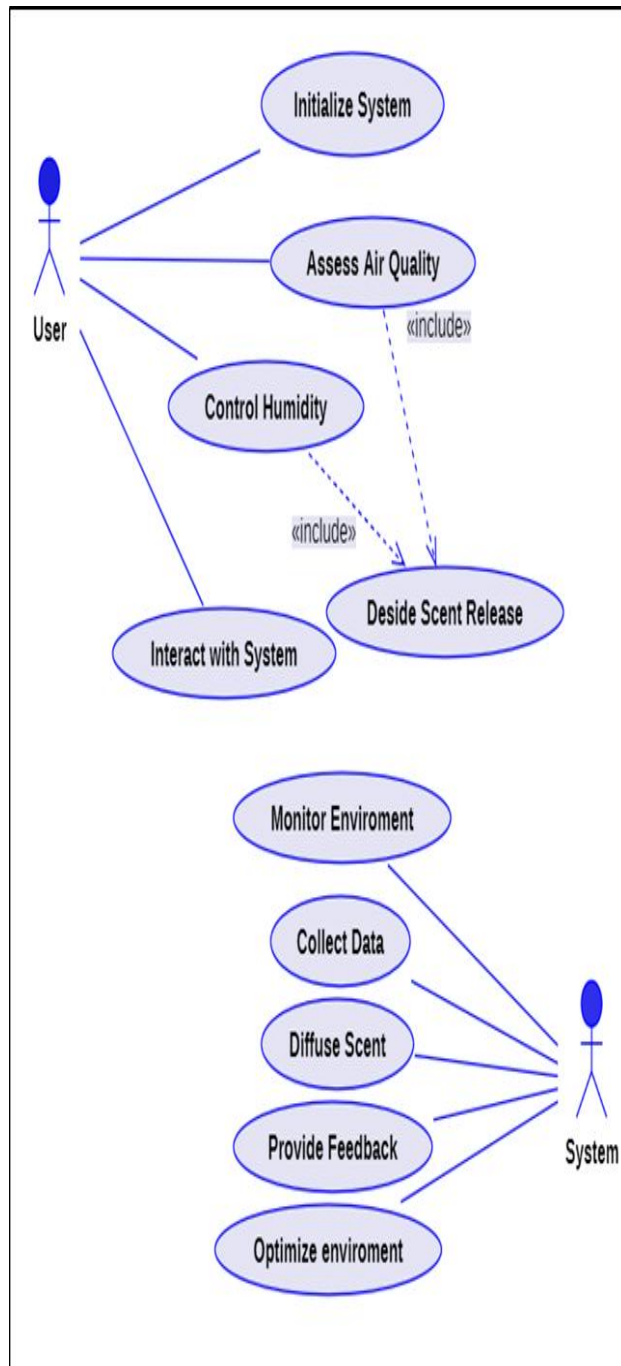


Figure 5: The use case scenario

Table 4 compares the applications of the Fragrance-Based Environmental Monitoring System across various environments. It outlines the primary goals, air quality monitoring strategies, and fragrance diffusion methods in different settings.

Table 4: Comparison of use cases applications in different environments

Aspect	Residenti al Living	Office Worksp ace	Healthc are Facilitie s	Hospita lity Industr y	Educatio nal Institutio ns
Primary Goal	Ensure a comfortable and refreshing atmosphere for family members	Optimize indoor environment for employee well-being and productivity	Maintain clean air to aid in patient recovery and comfort healthcare staff	Enhance visitor comfort and experience	Promote focus and comfort in the learning environment
Air Quality Monitor ing	Continuously assesses air quality to detect allergens or pollutants	Monitors air quality for dust and VOCs	Monitors air quality to ensure cleanliness	Maintains air quality by detecting smoking or other odors	Eliminates allergens and contaminants to ensure a clean environment
Fragran ce Diffusio n	Releases soothing fragrances to purify the air when pollutants are detected	Diffuses invigorating fragrances to rejuvenate the workspace when needed	Releases calming fragrances to soothe patients and staff	Disperses pleasant aromas to enhance guest experience	Uses mild fragrances to maintain focus and reduce dry air discomfort

6 Expected outcomes

The proposed Fragrance-Based Environmental Monitoring System improves indoor air quality (IAQ) through continuous monitoring and targeted fragrance diffusion. It enhances comfort through optimal humidity control and pleasant fragrance diffusion, improving physical and mental wellness. Real-time monitoring ensures ongoing assessment and corrective actions. Improved air quality reduces health risks, promoting better health for residents and workers, and enhancing productivity.

The expected performance of the system is illustrated in Figure 6, where real-time monitoring charts provide detailed insights into air quality and humidity levels, demonstrating the system's effectiveness in maintaining a healthy indoor environment.

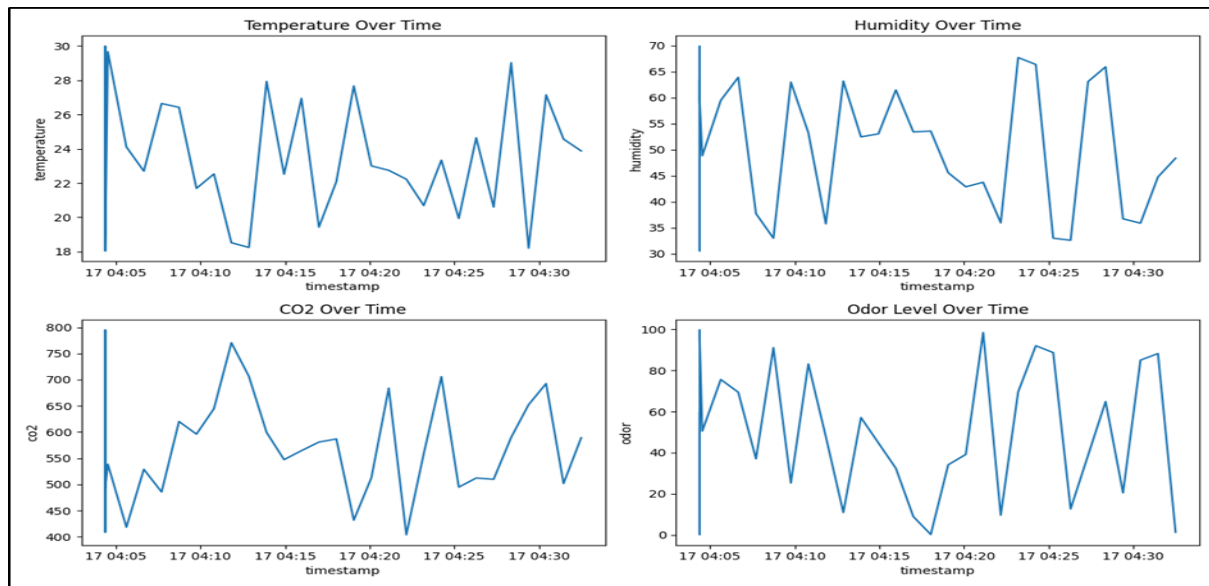


Figure 6: The real-time monitoring charts.

7 Conclusion

FBEMS represents an important advancement in intelligent air quality management. By integrating real-time monitoring with adaptive fragrance diffusion, the system ensures a sustainable and efficient approach to maintaining indoor air quality. This framework is expected to surpass conventional air purification methods by offering a cost-effective, energy-efficient, and user-centric solution.

Future research should focus on:

- Developing prototypes for real-world validation.
- Expanding system applications to various environments such as residential, commercial, and healthcare settings.
- Integrating additional AI-driven optimization techniques for enhanced system intelligence.

Acknowledgement

The authors thank Mustansiriyah University, (<https://uomustansiriyah.edu.iq>) in Baghdad, Iraq, for their support in the present work.

References

- [1] Fang, B., Xu, Q., Park, T., and Zhang, M.. AirSense: An intelligent home-based sensing system for indoor air quality analytics. *Proceedings of the 2016 ACM International Joint Conference on Pervasive and Ubiquitous Computing*, (pp. 109-119), 2016. <http://dx.doi.org/10.1145/2971648.2971720>
- [2] Pietraru, R.N., Olteanu, A., Adochiei, I.-R., and Adochiei, F.-C. Reengineering indoor air quality monitoring systems to improve end-user experience. 24(8), 2659, 2024. <https://doi.org/10.3390/s24082659>
- [3] AlSaad, S. N., and Hussien, N. M. Landmark-based shortest path detection in alarm systems. *Al-Mustansiriyah Journal of Science*, 29(2), 135-140, 2018. <https://doi.org/10.23851/mjs.v29i2.276>
- [4] Sun, S., Zheng, X., Villalba-Díez, J., and Ordieres-Meré, J. Indoor air-quality data-monitoring system: Long-term monitoring benefits. *Sensors*, 19, 4157, 2019. <https://doi.org/10.3390/s19194157>
- [5] Munir, F., Hakim, A., Hashim, S., and Iqbal, S. Air-Sense: Enhancing crop yield and quality through an integrated IoT-based air quality monitoring system. *Journal of Computing and Biomedical Informatics*, 6(02), 237-245, 2024. <https://doi.org/10.56979/602/2024>
- [6] Pant, A., Sharma, S., and Pant, K. Evaluation of Machine Learning Algorithms for Air Quality Index (AQI) Prediction. *Journal of Reliability and Statistical Studies*, 16(02), 229–242, 2023. <https://doi.org/10.13052/jrss0974-8024.1621>
- [7] Rollo, F., Bachechi, C., and Po, L. Anomaly detection and repairing for improving air quality monitoring. *Sensors*, 23(2), 640, 2023. <https://doi.org/10.3390/s23020640>
- [8] Brattoli, M., Loiotile, A. D., Lovascio, S., and Penza, M. Odour detection methods: Olfactometry and chemical sensors. *Sensors (Basel, Switzerland)*, 11(5), 5290-5322, 2011. <https://doi.org/10.3390/s110505290>
- [9] Moya, T. A., Ottelé, M., and Bluyssen, P. M. The effect of an active plant-based system on perceived air pollution. *International Journal of Environmental Research and Public Health*, 18(15), 2021. <https://doi.org/10.3390/ijerph18158233>
- [10] Paluchová, J., Berčík, J., and Horská, E. The sense of smell. In *Sensory and Aroma Marketing* (pp. 27-60). Wageningen Academic, 2017. https://doi.org/10.3920/978-90-8686-841-4_2
- [11] Steinemann, A. Ten questions concerning air fresheners and indoor built environments. *Building and Environment*, 111, 279-284, 2017. <https://doi.org/10.1016/j.buildenv.2016.11.009>

- [12] Marques, F. B., Bettoni, G. N., dos Santos, B. G. T., Adeoye, A. A., de Brito, B. G., de Brito, K. C. T. and Cavalli, L. S. AquaSafe: Aquaculture occupational safety and health in the palm of your hand. *Pesquisa Agropecuária Gaúcha*, 26(1), 46-54, 2020. <https://doi.org/10.36812/pag.202026146-54>
- [13] Patino, E. D. L., and Siegel, J. A. Indoor environmental quality in social housing: A literature review. *Building and Environment*, 131, 231-241, 2018.
<http://dx.doi.org/10.1016/j.buildenv.2018.01.013>
- [14] Hable-Khandekar, V., and Srinath, P. Machine learning techniques for air quality forecasting and study on real-time air quality monitoring. In 2017 International Conference on Computing, Communication, Control and Automation (ICCUBEA) (pp. 1-6). IEEE, 2017.
<http://dx.doi.org/10.1109/ICCUBEA.2017.8463746>
- [15] Younis, M. T., Hussien, N. M., Mohialden, Y. M., Raisian, K., Singh, P., and Joshi, K. Enhancement of ChatGPT using API wrapper techniques. *Al-Mustansiriyah Journal of Science*, 34(2), 82-86, 2023.
<https://doi.org/10.23851/mjs.v34i2.1350>
- [16] Saini, J., Dutta, M., and Marques, G. Machine Learning for Indoor Air Quality Assessment: A Systematic Review and Analysis. *Environmental Modeling and Assessment*, 1-18, 2024.
<http://dx.doi.org/10.1007/s10666-024-10001-1>
- [17] Saini, J., Dutta, M., & Marques, G. Sensors for indoor air quality monitoring and assessment through Internet of Things: a systematic review. *Environmental Monitoring and Assessment*, 193(2), 66, 2021.
<https://doi.org/10.1007/s10661-020-08781-6>
- [18] Zhang, H., & Srinivasan, R. A systematic review of air quality sensors, guidelines, and measurement studies for indoor air quality management. *Sustainability*, 12(21), 9045, 2020.
<https://doi.org/10.3390/su12219045>
- [19] E. A. W. Hachim, M. T. Gaata and T. Abbas. Iris-based Authentication Model in Cloud Environment (IAMCE). *International Conference on Electrical, Computer and Energy Technologies (ICECET)*, IEEE, 2022.
<http://dx.doi.org/10.1109/ICECET55527.2022.9873499>