INDOOR AND OUTDOOR AIR POLLUTION



- By: EmbedElites

TEAM 17

Team Members:

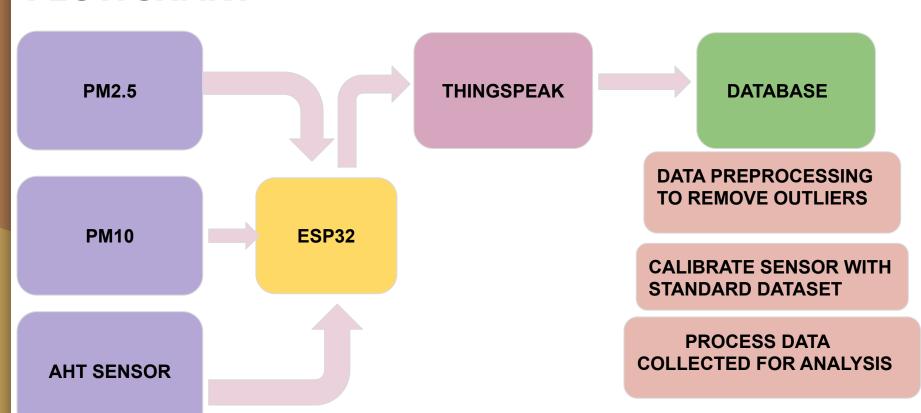
- 1. Archisha Panda
- 2. Paridhi Jain
- 3. Dharmineni Yashaswinee
- 4. Vyakhya Gupta

MOTIVATION

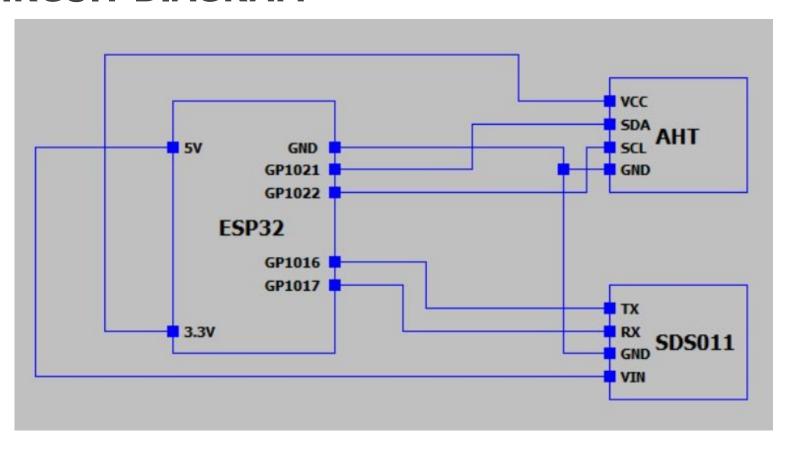
- **Objective**: Develop an 'Air Quality Monitor' to improve air pollution awareness.
- Parameters: Monitor PM 2.5, PM 10, Temperature, and Humidity levels.
- Caution System: The dashboard transforms AQI colors, serving as a cautionary indicator of pollution levels.
- **Deployment**: Install sensors indoors and outdoors
- Data Collection: Continuous data collection uploaded to ThingSpeak.
- Data Processing: Utilize machine learning for pollution analysis.
- Benefits: Plan pollution mitigation strategies and assess indoor and outdoor air quality



FLOWCHART



CIRCUIT DIAGRAM



TIMELINE

LAB1

Develop a comprehensive project plan that includes tasks, milestones, and dependencies.

LAB2

Acquire SDS011 and AHT sensors along with necessary components. Verified and understood the working of SDS011 and AHT sensor.

LAB7

PM 2.5 and PM 10 have been calibrated with reference to Aero PM standard dataset.

LAB8

Dashboard with basic amenities is created. Historical data, Real time data, Prediction of data can be accessed

LAB3

Hardware Integration: Assemble the sensors and interface them with a microcontroller (esp32). Started working on the zero PCB to build a deployable model

LAB6

Complete model has been built and deployed in the lab for 2 weeks data Sent almost 5000 entries to Thingspeak over this period

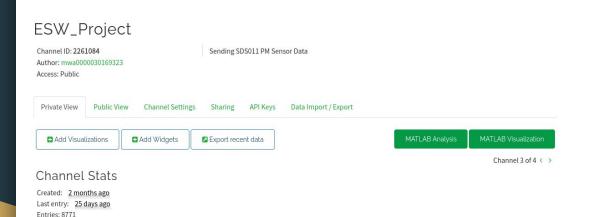
LAB5

Fetching data from ThingSpeak and pre-processing the data

LAB4

Data collection (>=2000 data points over 2 days) and sent to Thingspeak

DEPLOYMENT





The PCB constructed was deployed in NBH and our room with data entries collected and pushed to Thingspeak.

THINGSPEAK-CHANNEL 1 (INDOOR)

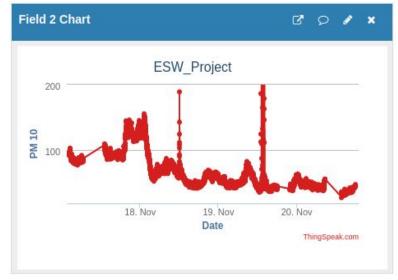
Channel Stats

Created: 2 months ago

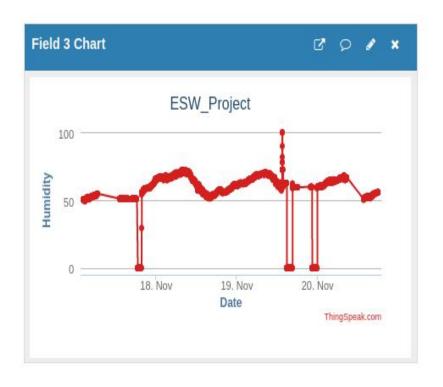
Last entry: about 9 hours ago

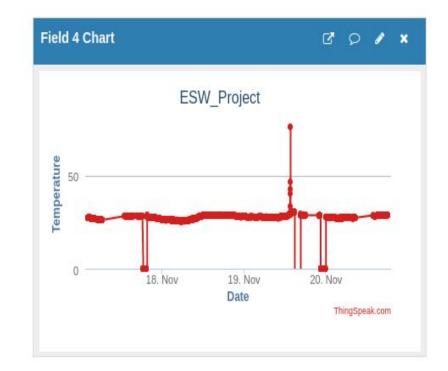
Entries: 14078





THINGSPEAK-CHANNEL 1 (INDOOR)





THINGSPEAK - CHANNEL 2 (OUTDOOR)

Channel Stats

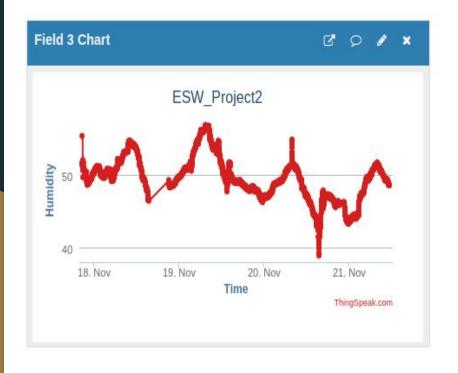
Created: about a month ago
Last entry: less than a minute ago

Entries: 10861





THINGSPEAK - CHANNEL 2 (OUTDOOR)





UPDATING DATABASE

The data is then fetched from ThingSpeak for pre-processing.

Standard		Standard	Standard	Standard	Standard
1 Timestamp		PM2.5	PM10	Humidity	Temperature
2 2023-10-03 6	93:33:35	31.40	59.90	87.58	26.41
3 2023-10-03 6	93:33:40	31.80	61.80	86.04	26.73
4 2023-10-03 6	33:33:45	31.80	62.10	85.32	26.60
5 2023-10-03 6	93:33:50	31.80	62.20	85.31	26.48
6 2023-10-03 6	93:33:55	32.10	64.30	85.43	26.38
7 2023-10-03 6	33:34:00	32.30	65.00	85.43	26.36
8 2023-10-03 6	93:34:05	32.20	65.60	85.36	26.31
9 2023-10-03 6	33:34:18	32.20	64.80	85.47	26.24
10 2023-10-03 6	33:34:20	32.50	64.80	85.78	26.18
11 2023-10-03 6	33:34:26	32.20	64.10	85.82	26.20
12 2023-10-03 6	33:34:33	32.20	63.90	85.81	26.18
13 2023-10-03 6	33:34:49	32.20	63.70	85.80	26.18
14 2023 - 10 - 03 6	33:34:56	31.70	62.20	85.90	26.19
15 2023-10-03 6	93:34:58	31.90	63.00	85.96	26.18
16 2023-10-03 6	93:35:02	31.60	60.70	86.01	26.16
17 2023-10-03 6	93:35:07	31.40	61.90	86.14	26.17
18 2023-10-03 6	93:35:11	31.20	59.80	86.16	26.18
19 2023-10-03 6	33:35:16	31.10	57.60	86.14	26.19
20 2023-10-03 6	33:35:22	31.00	58.20	86.09	26.22
21 2023-10-03 6	93:35:26	31.20	56.10	86.12	26.20
22 2023-10-03 6	93:35:32	31.10	56.20	86.15	26.20
23 2023-10-03 6	93:35:37	31.40	55.70	86.16	26.20
24 2023 - 10 - 03 6	93:35:41	31.80	56.70	86.24	26.19
25 2023 - 10 - 03	93:35:46	31.90	56.80	86.26	26.19
26 2023 - 10 - 03 6	93:35:52	32.40	57.50	86.22	26.16
27 2023 - 10 - 03	33:35:56	32.60	59.50	86.27	26.13
20 2023 10 03 0	3.36.03	22 80	60 30	96 40	26 00

DATA PREPROCESSING

- 1. Extract the relevant data from the JSON response using Read API key from ThingSpeak.
- 2. Remove duplicates according to timestamps.
- 3. Remove Outliers (eg. removing entries with RH>=90).
- 4. Interpolate to fill missing values.
- 5. Remove irrelevant data using interquartile range.
- 6. Retime to reduce the volume of dataset.
- 7. Maintain moving average for data consistency.
- 8. Raw data is preprocessed and ready for analysis.

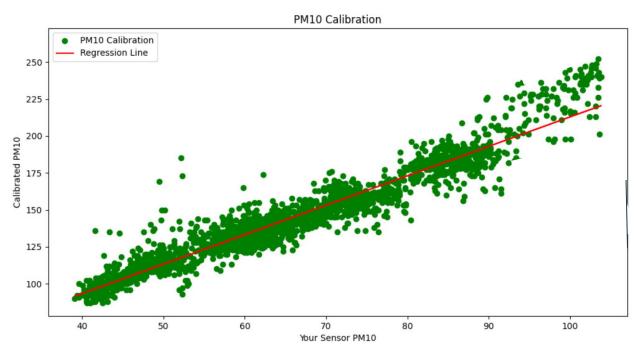
```
"channel": {
    "id": 2261084
    "name": "ESW Project",
    "description": "Sending SDS011 PM Sensor Data",
    "latitude": "0.0",
    "longitude": "0.0",
    "field1": "PM 2.5
    "field4": "Temperature",
    "created at": "2023-09-05T09:27:59Z",
    "updated_at": "2023-09-26T09:57:41Z",
    "last entry id": 5458
},
"feeds": [
        "created at": "2023-10-03T04:43:07Z",
        "entry id": 5359,
        "field1": "27.20",
        "field2": "60.00",
        "field3": "74.32",
        "field4": "27.28"
        "created_at": "2023-10-03T04:43:12Z",
        "entry id": 5360,
        "field1": "28.20"
        "field2": "58.60"
        "field3": "74.87",
        "field4": "27.15"
        "created at": "2023-10-03T04:43:17Z",
        "entry_id": 5361,
        "field1": "28.80".
        "field2": "57.80",
        "field3": "75.69"
        "field4": "27.07"
        "created at": "2023-10-03T04:43:23Z",
        "entry id": 5362,
        "field1": "28.70",
        "field3": "76.05"
        "field4": "26.93"
        "created at": "2023-10-03T04:43:27Z",
        "entry id": 5363,
        "fieldī": "29.10"
        "field2": "61.00"
        "field3": "76.35"
```

- We were provided with a standard dataset collected from Aero PM sensor.
- Data from our database was pre-processed to remove any outliers and ensure consistency in data points.
- The pre-processed data and standard dataset was then used for calibration of sensor.

Date Time	Monitor ID	Location ID	pm_10	pm_25
13 Oct 2023 22:44	2	1	0.129	0.064
13 Oct 2023 22:45	2	1	0.127	0.064
13 Oct 2023 22:46	2	1	0.139	0.067
13 Oct 2023 22:47	2	1	0.13	0.063
13 Oct 2023 22:48	2	1	0.138	0.067
13 Oct 2023 22:49	2	1	0.13	0.065
13 Oct 2023 22:50	2	1	0.13	0.064
13 Oct 2023 22:51	2	1	0.13	0.065
13 Oct 2023 22:52	2	1	0.134	0.066
13 Oct 2023 22:53	2	1	0.134	0.067
13 Oct 2023 22:54	2	1	0.142	0.069
13 Oct 2023 22:55	2	1	0.132	0.062
13 Oct 2023 22:56	2	1	0.132	0.066
13 Oct 2023 22:57	2	1	0.128	0.064
13 Oct 2023 22:58	2	1	0.127	0.066
13 Oct 2023 22:59	2	1	0.135	0.066
13 Oct 2023 23:00	2	1	0.13	0.066
13 Oct 2023 23:01	2	1	0.136	0.067
13 Oct 2023 23:02	2	1	0.137	0.067
13 Oct 2023 23:03	2	1	0.127	0.064
13 Oct 2023 23:04	2	1	0.138	0.068
13 Oct 2023 23:05	2	1	0.135	0.067
13 Oct 2023 23:06	2	1	0.141	0.067
13 Oct 2023 23:07	2	1	0.136	0.067
13 Oct 2023 23:08	2	1	0.132	0.067
13 Oct 2023 23:09	2	1	0.14	0.067
13 Oct 2023 23:10	2	1	0.128	0.064
13 Oct 2023 23:11	2	1	0.128	0.067
13 Oct 2023 23:12	2	1	0.138	0.066

AERO PM DATA

- The data points are aligned according to timestamps.
- Plot local data versus standard dataset and obtain a linear relation between them (y=mx+c where y: standard dataset, x: local data, m:scaling factor and c: offset)



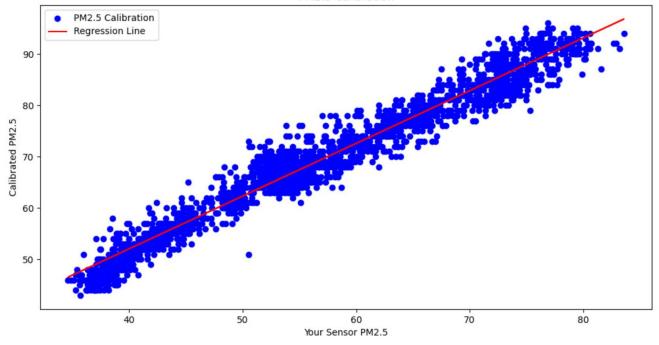
Result:

Scaling Factor (PM10): 1.9935275505072367

RMSE (PM10): 9.665478354202433

Offset (PM10): 13.568800635884315

PM2.5 Calibration



Result:

Scaling Factor (PM2.5): 1.0256618336813879

RMSE (PM2.5): 2.7046325971438927

Offset (PM2.5): 11.048211689089669

Now we can plug our collected data points into this equation to obtain legitimate data for analysis.

Coefficient of Determination:

R-squared (PM10): 0.9098807762004294 R-squared (PM2.5): 0.9547599726165852

An R-squared value above 0.7 is generally considered a strong fit. It suggests that the calibration explains a large proportion of the variance in the data, and it is providing useful information for prediction or analysis.

Our R-squared obtained using linear regression for SDS011 sensor $\sim 0.9 > 0.7$. Hence, the sensor is now calibrated to give correct readings.

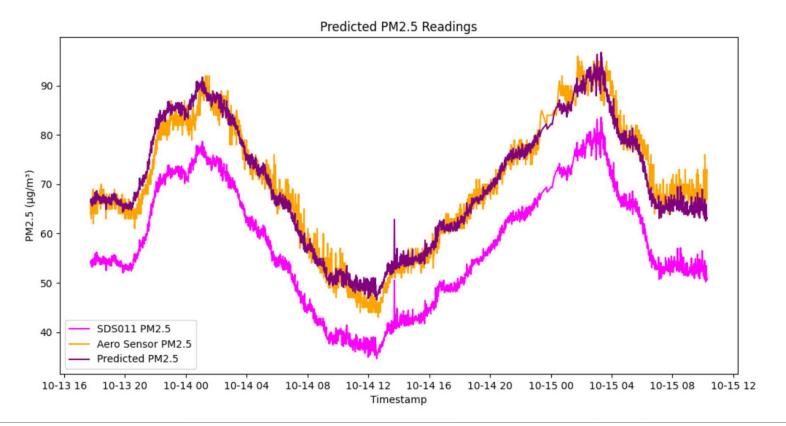
Formula

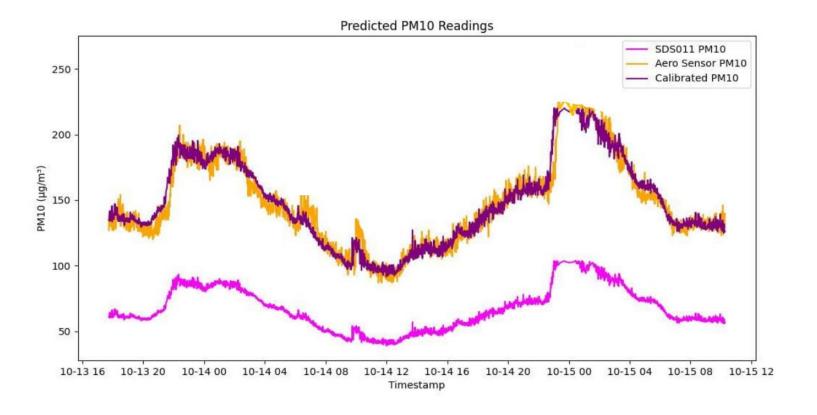
$$R^2 = 1 - \frac{RSS}{TSS}$$

 R^2 = coefficient of determination

RSS = sum of squares of residuals

TSS = total sum of squares





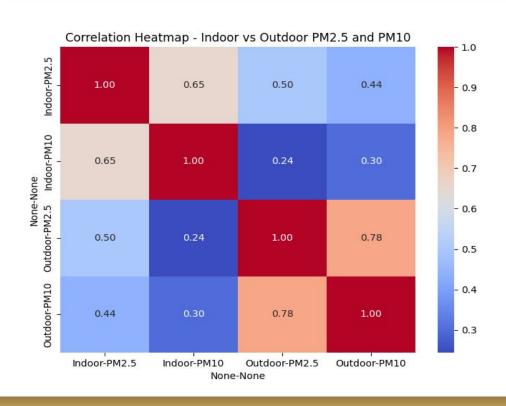
COEFFICIENT OF VARIABILITY OF DATA (INDOOR)

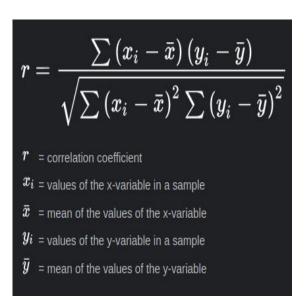
Coefficient of variation is a relative measure of dispersion that is used to determine the variability of data.

$$CV=rac{\sigma}{\mu}$$
 σ = population standard deviation μ = population mean

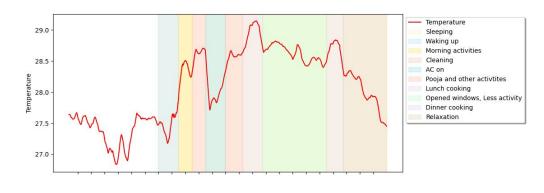
Coefficient of Variation (CV) for PM2.5 data: 23.95% Coefficient of Variation (CV) for PM10 data: 24.78% Coefficient of Variation (CV) for Humidity data: 11.78% Coefficient of Variation (CV) for Temperature data: 12.67%

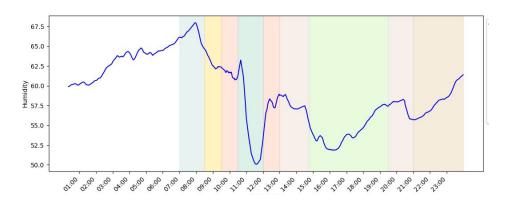
CORRELATION BETWEEN INDOOR AND OUTDOOR PM





ANALYSIS

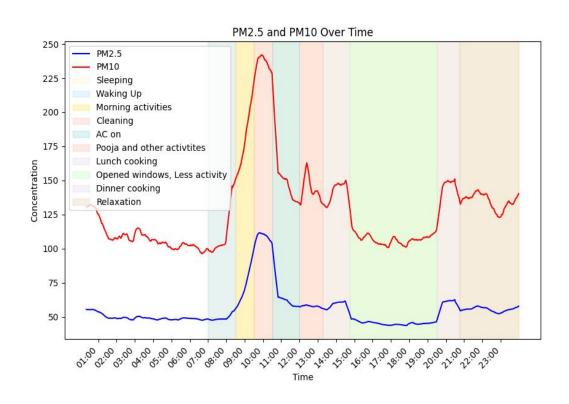




Analyzing the graph from this slide reveals fluctuations in temperature and humidity during different activities in an Indian household:

- Morning wake up shows a temperature drop and humidity rise.
- Daytime and cleaning exhibit rising temperature and declining humidity.
- Pooja and cooking result in a temperature hike with minimal humidity change.
- Opening windows leads to decreased humidity and maintained temperature.
- Nighttime sees a temperature decrease and a slight humidity rise.

ANALYSIS

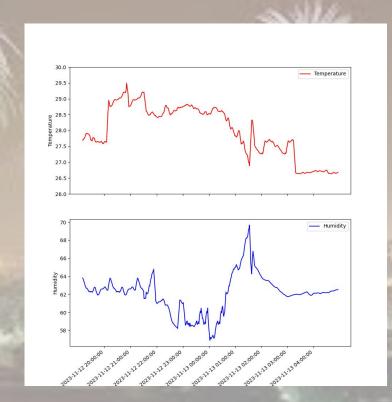


- Analyzing the PM10 and PM2.5 graph indicates minimal pollution during sleep.
- Upon waking and engaging in morning activities and cleaning, pollution levels gradually rise, spiking notably during cleaning due to dust particles.
- AC usage maintains low PM10 and PM2.5 readings.
- During puja and cooking, a slight increase in PM10, with a similar trend during dinner cooking.
- Levels decrease during less activity and moderately increase during relaxation at night.

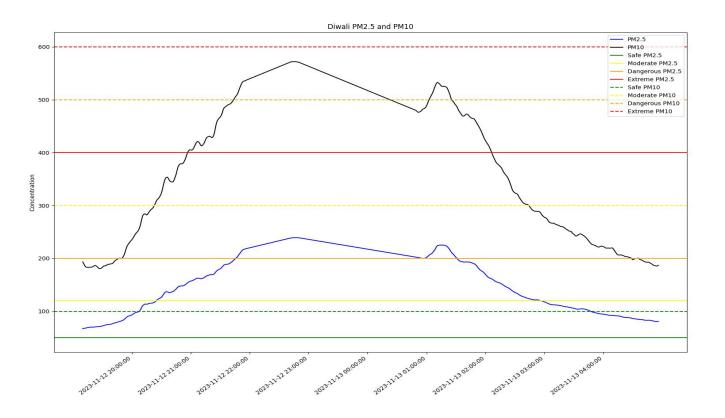
OBSERVATIONS

Diwali

- During Diwali, due to the bursting of crackers, there was a noticeable surge in pollution levels.
- This increase in pollution levels was evident in the higher concentrations of particulate matter, with PM2.5 registering at 231.57 and PM10 at 579.73.
- The festive activities contributed to a rise in humidity, recorded at 66.43%, and a temperature increase to 27.99°C



OBSERVATIONS



OBSERVATIONS

1. Cleaning

While cleaning, we noticed a rise in PM2.5 and PM10

levels. This increase is likely due to the disturbance of settled dust, use of cleaning agents, and mechanical actions involved in the cleaning process.

2. Boiling

Boiling activities result in a simultaneous elevation of temperature and humidity levels, noticeable during the process.

3. <u>Incense Stick</u>

The burning of incense sticks contributes to an escalation in PM2.5 and PM10 levels, influencing air quality during their combustion.





Home Indoor Pollution Outdoor Pollution

Welcome to the Air Pollution Dashboard

Welcome to our Air Pollution Monitoring Dashboard, your go-to resource for real-time and historical data on air quality in your area.

Our mission is to empower you with the knowledge you need to make informed decisions about your daily activities, health, and well-being.

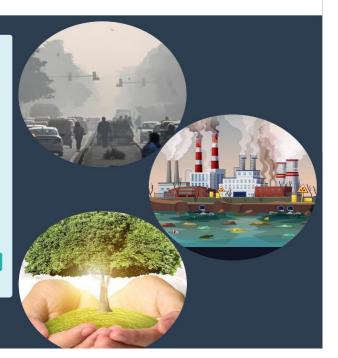
Air quality affects us all, and our interactive dashboard provides up-to-the-minute Air Quality Index (AQI) readings, pollutant concentrations, and visualizations that help you understand the state of the air in your region.

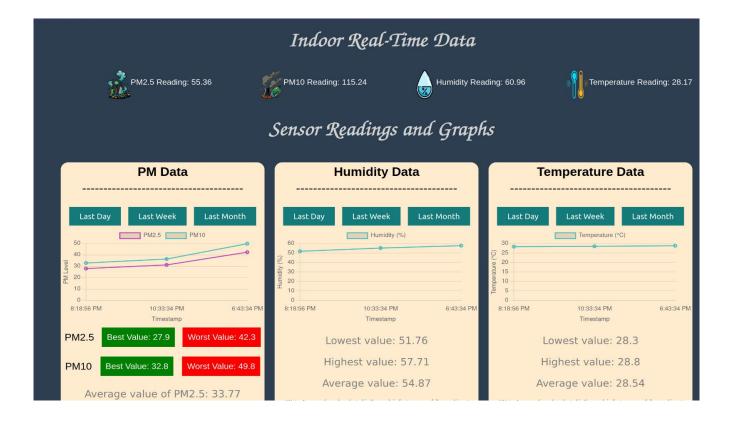
Whether you're concerned about outdoor exercise, planning a commute, or simply curious about the air you breathe, our user-friendly platform delivers the information you need.

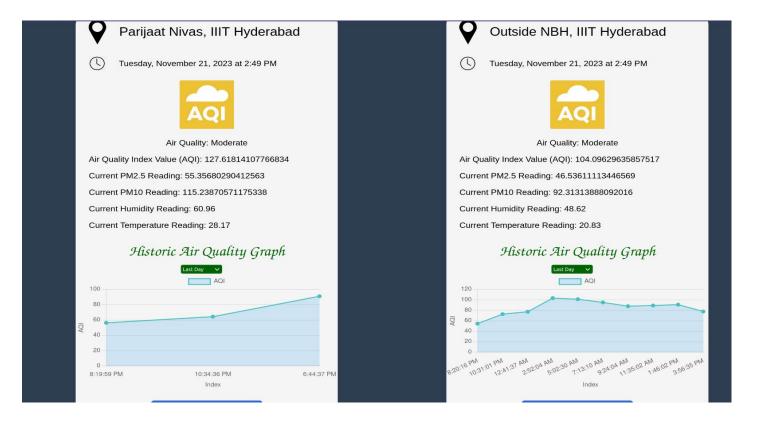
Explore our maps, charts, and resources, and stay connected to the latest updates and alerts.

We're dedicated to fostering a cleaner and healthier environment, one informed choice at a time.

Click to Know More >>







The dashboard will show:

HOME:

- Navigates to other components of dashboard
- Description of our project
- Shows air quality of indoors and outdoors
- Current readings of all sensors for indoors and outdoors

REALTIME: PM 2.5, PM 10, Temperature, Humidity data of real time is collected from sensors, published to thingspeak and exhibited here

HISTORICAL ANALYSIS: Data retrieved from thingspeak of last day, last week, last month is shown here.

CALCULATION OF AIR QUALITY INDEX (AQI)

For PM2.5:

$$AQI_{PM2.5} = rac{AQI_{high} - AQI_{low}}{Conc_{high} - Conc_{low}} imes (Conc_{PM2.5} - Conc_{low}) + AQI_{low}$$

For PM10:

$$AQI_{PM10} = rac{AQI_{high} - AQI_{low}}{Conc_{high} - Conc_{low}} imes (Conc_{PM10} - Conc_{low}) + AQI_{low}$$

Where:

- ullet AQI_{high} and AQI_{low} are the AQI breakpoints for PM2.5 and PM10, respectively.
- ullet $Conc_{high}$ and $Conc_{low}$ are the corresponding concentration breakpoints for PM2.5 and PM10, respectively.
- ullet $Conc_{PM2.5}$ and $Conc_{PM10}$ are the actual measured concentrations of PM2.5 and PM10, respectively.

CONCLUSION

- 1. Data collection done by us during Diwali highlighted the significant rise in pollution levels. This could be captured even by sensors deployed inside our room which is alarming. The PM2.5 and PM10 values shooted to 231.57 and 579.73 respectively which is highly unsafe according to health standards.
- 2. For indoor pollution, we performed several activities like cooking, dusting, etc over the span of a day to simulate an average Indian household. The data collected showed variations in PM2.5 and PM10 and other parameters associated with these activities indicating extent to which even indoor pollution can impact our health.
- 3. Combining the results of indoor and outdoor data collection, we concluded that the correlation factor of a room not very well ventilated to outdoors is less than the correlation factor a well-ventilated room would achieve.

CHALLENGES FACED

- Outdoor sensor deployment on Friday limited data points for initial analysis.
- Frequent disconnection due to power shortages and network issues results in missing data.

