Practical Approaches to IAQ Monitoring: From Sensor Selection to Data-Driven Decision Making





Dr. Julien Stamatakis Attune (formerly Senseware) julien@attuneiot.com

Agenda

- Indoor Air Quality Monitoring Basics
- Indoor Air Quality Sensors, Hardware & Communications
- IAQ Monitoring User Case
- Interactive Activity
- The Future of IAQ Monitoring



Learning Objectives

- **1. Understand the Fundamentals of IAQ Monitoring:** Learn the basic principles of IAQ monitoring and how it impacts building environments and occupant health.
- 2. Evaluate and Select Appropriate IAQ Sensors: Gain knowledge about the different types of IAQ sensors available, including their strengths, weaknesses, and use cases.
- **3.** Analyze and Interpret IAQ Data: Learn how to read and interpret IAQ data to make informed decisions about building operations and air quality improvements.
- **4.** Apply IAQ Data in Problem Solving: Participate in a hands-on exercise to identify IAQ concerns from real data, applying your knowledge to diagnose potential sources of air quality issues.
- **5. Stay Current with Emerging Sensor Technologies:** Explore cutting-edge IAQ sensor technologies and how they can be integrated into future-proof monitoring systems.



What is Indoor Air Quality (IAQ) Monitoring?

An IAQ monitor is a device or system that checks the levels of different air contaminants and helps assess how safe the air is to breathe. It can alert people when some of those levels become unsafe.

Bad air quality can affect our comfort, our health, and our cognitive performance.

The COVID-19 pandemic and climate change (eg, wildfires and increased ambient air pollution events) have highlighted the importance of good ventilation and filtration for good indoor air quality.

Air quality monitoring is critical to ensure that IAQ mitigation techniques are working.

The rise of Internet of Things and cloud technology has revolutionized the world of IAQ Monitoring.

After COVID, many companies released IAQ products. It is important to know how to navigate the noise to make sure you use the right product for what you need.

Let's Dive In.



What can we monitor and why?

Temperature and Humidity levels affect comfort and health. Too high humidity can promote mold growth, dust mites, and bacterial proliferation, while too low humidity can cause respiratory irritation and dry skin.

High **Carbon Dioxide** levels can indicate poor ventilation and overcrowding. Elevated CO₂ levels can indicate underventilation which can cause drowsiness, headaches, and reduced cognitive performance. Low CO₂ can indicate overventilation.

Volatile Organic Compounds (VOCs) are emitted as gases from certain solids or liquids, including paints, cleaning supplies, and adhesives. High levels can cause eye, nose, and throat irritation, headaches, and even long-term health issues such as liver or kidney damage.

Formaldehyde (HCHO) is a specific type of VOC commonly found in building materials and household products. It can cause respiratory problems and is classified as a human carcinogen.

Particulate matter can include dust, pollen, soot, and smoke. High concentrations can exacerbate respiratory issues, cardiovascular diseases, and other health problems, especially in vulnerable populations like children and the elderly.

While beneficial in the upper atmosphere, Ozone at ground level can cause respiratory problems and other health issues.

Carbon Monoxide (CO) is a colorless, odorless gas produced by burning fuel. High levels can lead to poisoning, causing symptoms ranging from headaches and dizziness to more severe outcomes like loss of consciousness or death.

And Many Others, including **Nitrogen Dioxide**. Produced by combustion processes, NO₂ can irritate airways, reduce lung function, and increase susceptibility to respiratory infections.



Sensors, Hardware & Communications

Hardware: Types of Monitors



Reference instruments

Expensive Very Accurate For Reference & Calibration



Handheld instruments

Cheaper **Point in time** Local access **Low Cost Sensors**



Real-time monitoring

Low-Cost **Real-time** Access from anywhere



IAQ Monitors: Wired vs Wireless



Wired Output

0-5V, 4-20mA, Modbus, BACnet, etc. Cheaper Device Expensive Installation **Local Access**

No Security Limited # of Sensors

Screen/App

70.275 65

aranete

Bluetooth No subscription **Consumer Type**



Wireless Output

WiFi, ZigBee, 5G, etc. More Expensive Device Cheaper Installation Encrypted Communications **Cloud Access** Unlimited data

IAQ Monitors: Types of Wireless





WiFi

Easy to Set Up **Difficult to Maintain** Medium Range Cheap High Bandwidth Easy to Set Up Easy to Maintain Long Range **Expensive** High Bandwidth

Cell



802.15.4

Easy to Set Up Easy to Maintain **Medium Range** Cheap Medium Bandwidth



Bluetooth

Difficult to Set Up

Difficult to Maintain Small Range Cheap High Bandwidth



LoRa

Difficult to Set Up Easy to Maintain Long Range Cheap **Very Low Bandwidth**



9

From the Sensors to an App





Example

Aranet 4



Good for short term monitoring



From the Sensors to the Cloud





Example

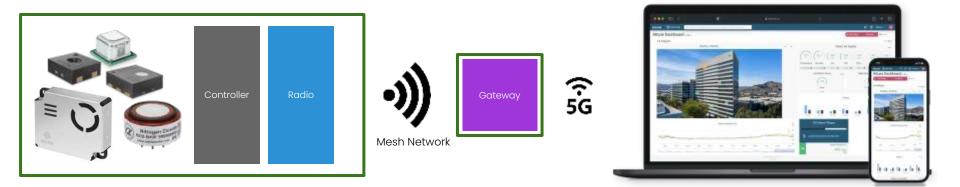
Good for small deployments

Wilfi

Purple Air, Kaiterra, Airthings



From the Sensors to the Cloud



Examples

Good for large deployments



Attune, Awair

Hardware: Types of Sensor Data





Temperature Relative Humidity

CO2

Small Very Accurate Low Drift Small NDIR Or Acoustic (smaller but less accurate)



Hardware: Types of Sensor Data – CO₂





Very Accurate More Expensive

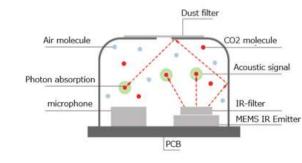
Ammonia

Mathematic

Carbon distate

NDIR (Non-Dispersive Infrared absorption spectroscopy)





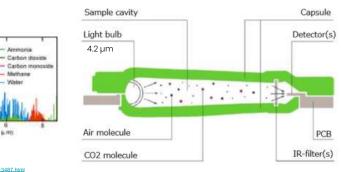
Smaller Cheaper Less Accurate

Not suited for outdoor use? (Wider range of T and RH values)





Wavelength & m)



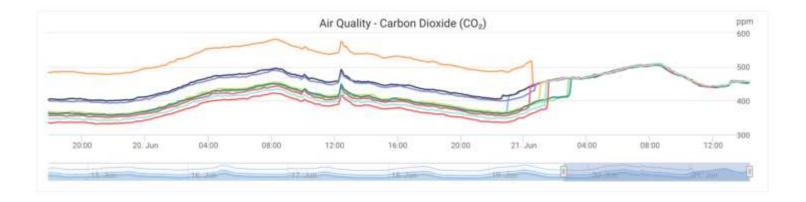
ot.com/blog-13487.htm

ttps://www.airgradient.com/blog/co2-sensors-photo-acoustic-vs-ndir-updated/ ttps://sensition.com/media/documents/8E7D4521/627C2C8D/CD_IN_SCDvv_1

and photoacoustic NDIR sensing D1.pdf

Hardware: Types of Sensor Data – CO₂

Sensors drift over time, Automatic Baseline Correction (ABC) can be used to maintain sensor calibration, **assuming** the sensor reaches 400ppm fresh air levels at least once a week.





Hardware: Types of Sensor Data







Temperature Relative Humidity CO_2

тиос

Small Very Accurate Low Drift Small NDIR Or Acoustic (smaller but less accurate) Small MOX Technology **Evolving Rapidly**

Other options: PID



Hardware: Types of Sensor Data – TVOC

Metal Oxide Sensors (MOX)

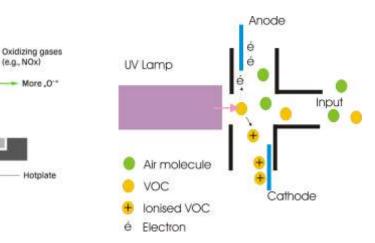
Surface

oxygen

0

Ω

MOX resistance signal



PID VOC sensors can help narrow down types of VOCs but require frequent calibration

Different Lamp Energy (eV) can be used to target specific VOCs

Photoionization (PID)



https://sensirion.com/media/documents/0083CDF4/6294DFEA/Info_Note_MOX_sensor.pdf https://sysmatec.ch/en/voc-measurement-with-pid-sensor/ https://www.engineerlive.com/content/how-do-you-choose-voc-sensor

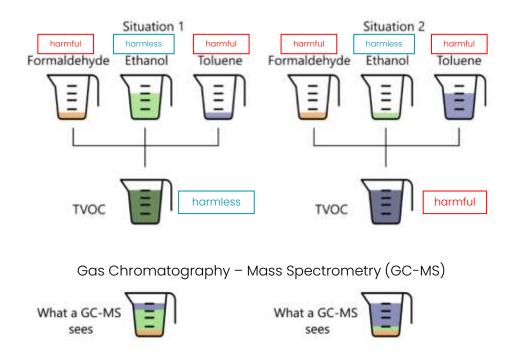
Reducing gases

(e.g., VOCs)

Less "O"

MOX layer

Hardware: Types of Sensor Data – TVOC



MOX TVOC sensors cannot discriminate between VOC types

AI MOX TVOC show some promise of VOC type classification



Hardware: Types of Sensor Data



Temperature Relative Humidity

CO2

тиос

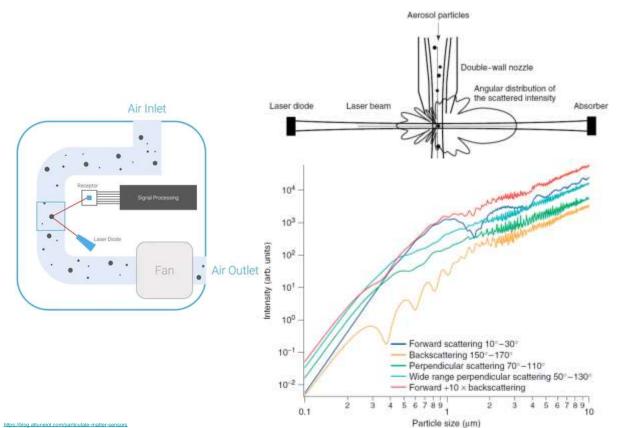
Particulate Matter

Small Very Accurate Low Drift Small NDIR Or Acoustic (smaller but less accurate) Small MOX Technology **Evolving Rapidly** Light Scattering Most difference between devices

Other options: PID



Hardware: Types of Sensor Data – PM



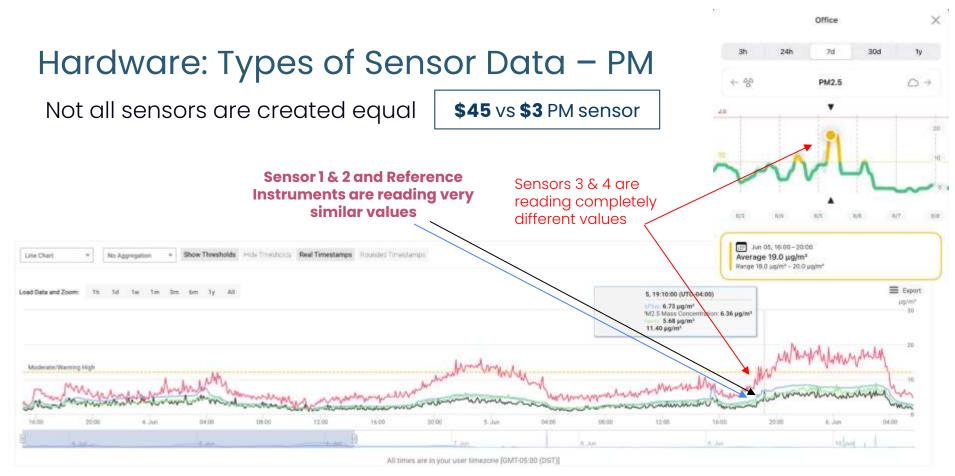
Important Factors

Airflow [can it bring in PM10 particles?]

Scattering Angle [can it measure size accurately?]

Humidity Compensation [does it represent true distribution?]







Hardware: Types of Sensor Data – PM

| PM Sensors | | | | | | | | |
|-----------------|-----------------|--------------------|-------------------|-----------------------|---------------------|--------------------|------------------|-------------------|
| Sensor Image | Make (Model) | Est. Cost (USD) | Pollutant(s) | *Field R ² | "Lab R ² | *Field MAE (µg/m³) | *Lab MAE (µg/m³) | Summary Report |
| Device | | \$5,000 | PM _{2.5} | 0.66 to 0.81 | 0.99 | 2.9 to 5.1 | 3.0 to 13.3 | PDF (723 KB) |
| | | | PM _{1.0} | 0.68 to 0.70 | | 2.4 to 2.5 | | |
| Device | | \$1,000 | PM _{2.5} | 0.54 to 0.57 | | 4.8 to 5.0 | | |
| | | | PM10 | 0.03 to 0.05 | | 19.7 to 19.8 | | |
| I | Device | \$300 | PM _{2.5} | ~ 0.0 | | 9.5 to 17.8 | | |
| | | nsor \$100 | PM _{1.0} | 0.91 | 0.99 | 1.3 to 1.4 | 0.8 to 1.4 | PDF (889 KB) |
| | Sensor | | PM _{2.5} | 0.80 to 0.83 | 0.99 | 2.0 to 5.1 | 5.4 to 6.5 | |
| | | | PM10 | 0.07 to 0.20 | + | 10.8 to 24.7 | | |

http://www.aqmd.gov/aq-spec/evaluations/criteria-pollutants/summary-pm



Hardware: Types of Sensor Data











Temperature Relative Humidity CO2

тиос

Particulate Matter

Special Gases

Small Very Accurate Low Drift Small NDIR Or Acoustic (smaller but less accurate)

Small MOX Technology **Evolving Rapidly** Light Scattering Most difference between devices Formaldehyde Ozone CO, NO2, SO2, etc.



Other options: PID

A Few Thoughts

- **Sensors:** Use 1 Type of instrument for a study or for Indoor/Outdoor comparison
- Sensor Accuracy: Remember the trade offs you made during data collection when you analyze the data
- Sensor Placement and Coverage: Proper placement of sensors is crucial for accurate monitoring. Sensors should be placed at representative locations to capture a true picture of the indoor environment.
- **Environmental Factors:** Consider how temperature, humidity, and other environmental conditions might affect sensor readings. Some sensors may require specific conditions to function optimally.

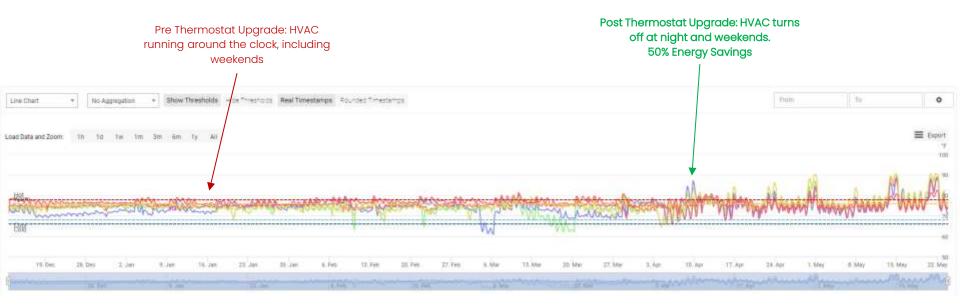


Use Case

With Building Data, you can't manage what you don't measure

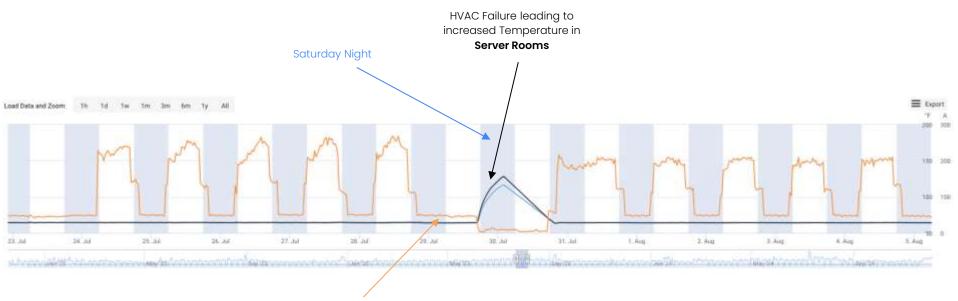


Temperature and Humidity – HVAC Issues





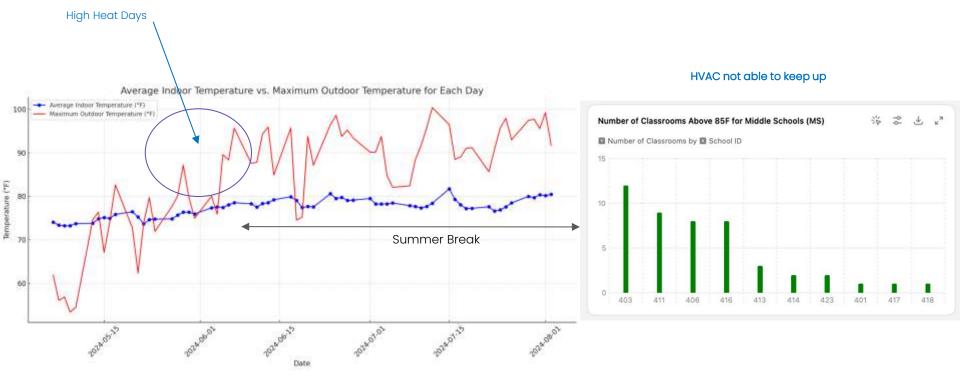
Temperature and Humidity – HVAC Failures



kW Data for Comparison

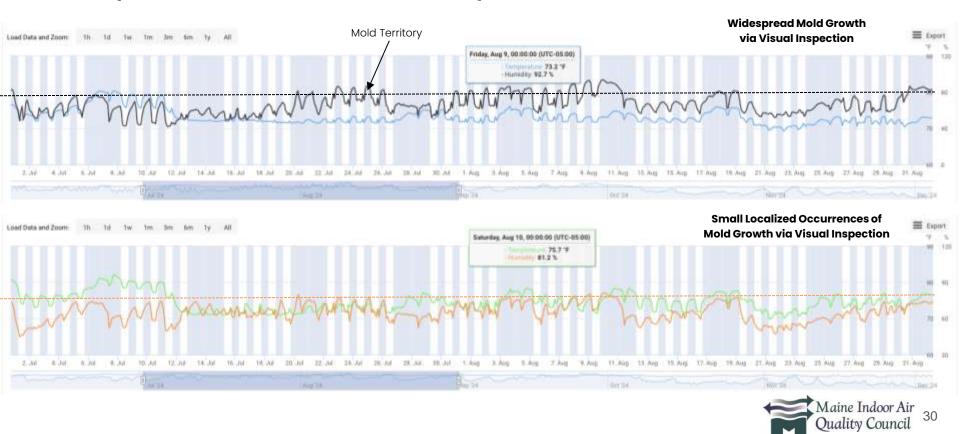


Temperature and Humidity – High Heat Days

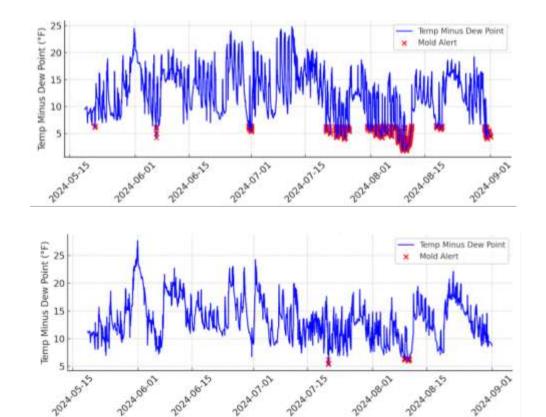




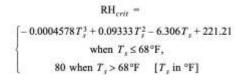
Temperature and Humidity – Mold Detection



Temperature and Humidity – Mold Detection



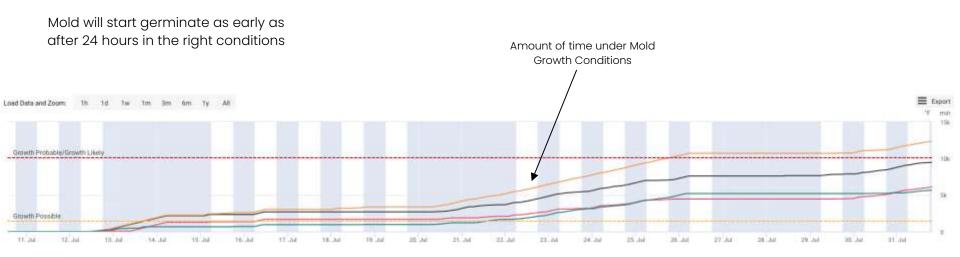
Widespread Mold Growth via Visual Inspection



Small Localized Occurrences of Mold Growth via Visual Inspection

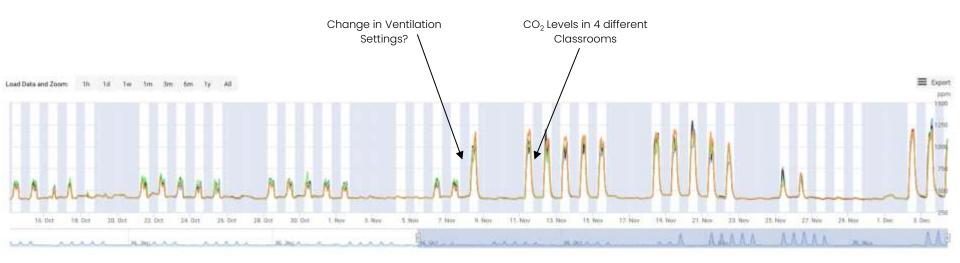


Temperature and Humidity – Mold Detection





Carbon Dioxide (CO₂) – Ventilation Levels





Carbon Dioxide (CO_2) – Ventilation Rates

eACH:

Data to be analyzed when there are no CO2 sources in the room

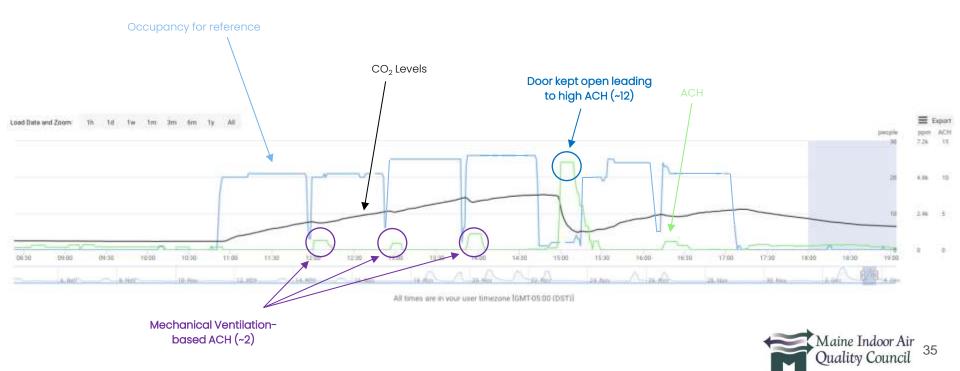
 Helpful to take into account natural ventilation, e.g. doors or window open

Calculations based on CO_2 in IAQ ACH = $1/\Delta t \ln[(C1-CR)/(C0-CR)]$

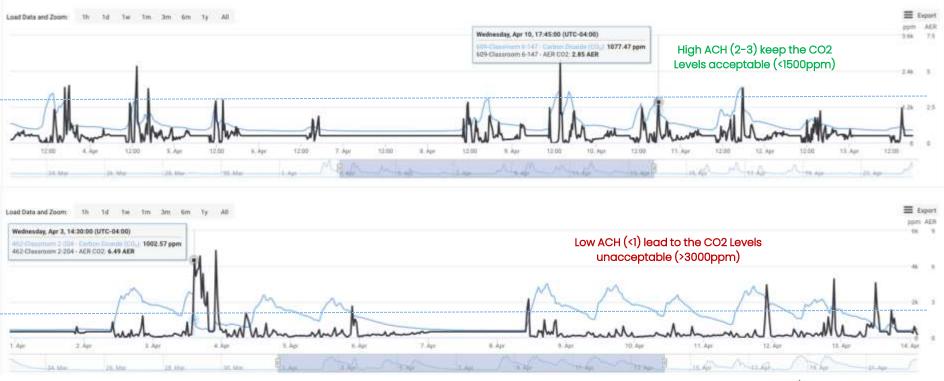




Carbon Dioxide (CO_2) – Ventilation Rates



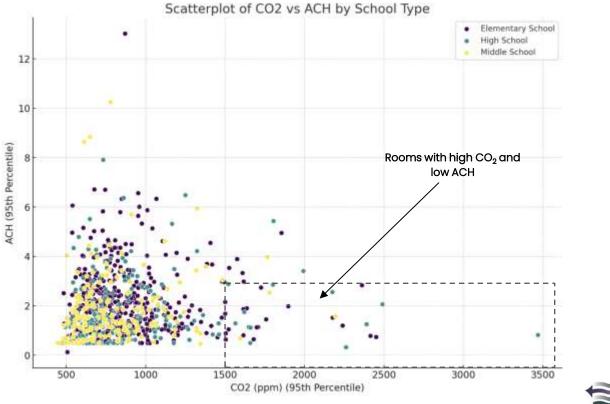
Carbon Dioxide (CO_2) – Ventilation Rates





36

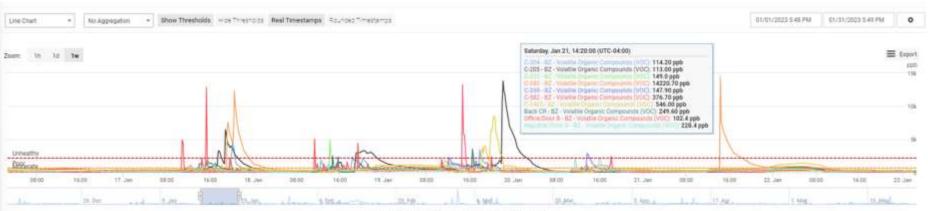
Carbon Dioxide (CO_2) – Ventilation Rates



Maine Indoor Air Quality Council 37

Total Volatile Organic Compounds (TVOC)

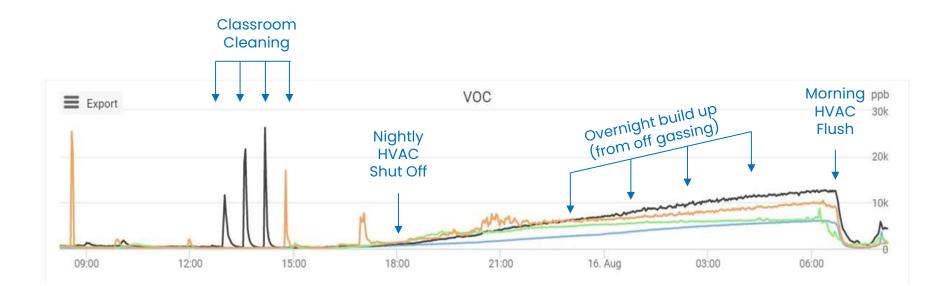
Continuously elevated levels of TVOC across multiple classrooms highlighted poor ventilation and over-use of cleaning products. This triggered a HVAC retrofit project to improve Ventilation, IAQ, Energy Efficiency, and Cleaning Protocols



AS times are in your user timezone (GMT-05.00 (DST))

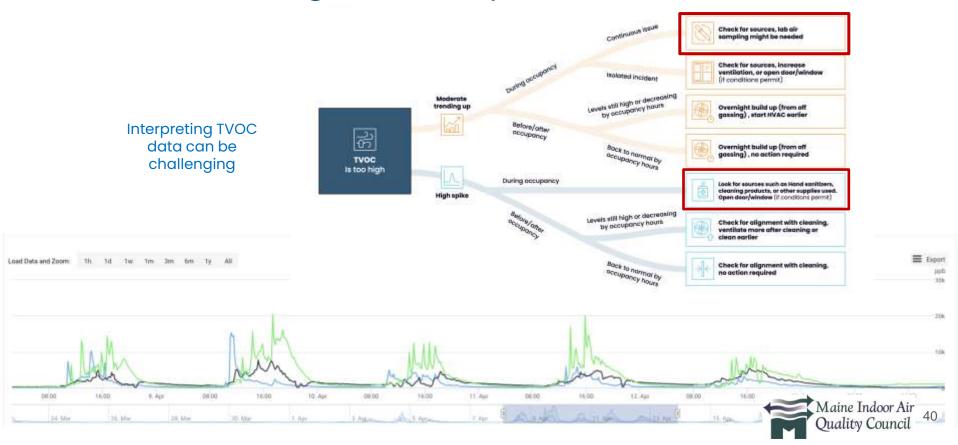


Total Volatile Organic Compounds (TVOC) - Events





Total Volatile Organic Compounds (TVOC) - Events



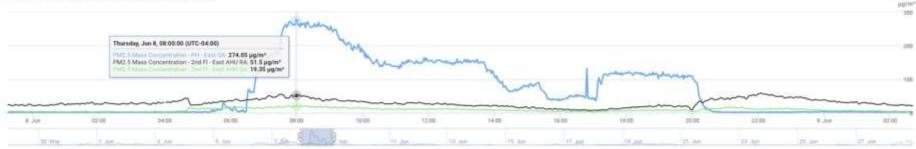
Particulate Matter (PM) – Wild Fires

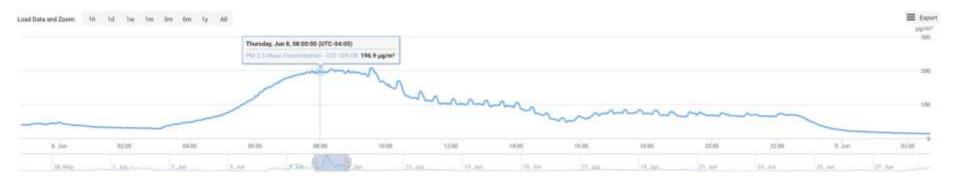
| | fee Dart * [12m * Anny Mari * Shire Labels Hite Labels Film Labels | <u>j</u> .n | | | |
|-------|--|--|------------------|--|--|
| fice | Local Data and Zoure 10 10 10 too line line data 10 AB | | = | | |
| ling | | 23.79ug/w ⁴ 32.41µg/w ⁴ | | | |
| • | Monthay, Jul S, 00.06.00 (UTC-04.00) Segment 18.15 app/mit AND 2 West Explor Are Sensor Charter PM 2.5 Mour Commission 0.36 pag/mit 18.16 pag/mit | 21.30eg/w ¹ | flugiw' | | |
| 61 | B.17pp/m* 2.55pp/m* 3.75pp/m* 2.75pp/m* 8.75pp/m* 3.75pp/m* <t< td=""><td>ant a stream</td><td>e Stygen*</td></t<> | ant a stream | e Stygen* | | |
| ire | | 10 A.M. | 1 | | |
| mpact | All times are in your user timesters (0x45.00.00 (0001)) | | | | |
| | Ber Chart * 13ar * Average (Muni) * Other Labels Hide Labels | 1 | | | |
| | Land Data and Zaany 10 tol two ton ton toy ad | | Ŧ | | |
| ol | | 25.76pg/m* | e : | | |
| | Mendey, Jul 1, 08:80:09 (UTC-64:69) Manifer, Tel 1 (and 100) Mill 1 Man Consentation Mari Tension Mari Tension Mari Tension Mari Tension Altington Altington Altington Altington Altington Altington Altington Altington Altington Altington Altington Altington Altington | 71.50pg/m ² 20 skpg 12.00pg/m ² | ine ² | | |
| | Litegine' Litegi | | 11 | | |
| | the last and the last the second second the second | 10.00 | | | |
| | All Virian ann W anar waard timeacana IOMT 05:00 (2017) | | | | |



Particulate Matter (PM) – Wild Fires

Load Data and Zoom. 1h 1d 1w 1m 3m 6m 1y All







Export

Particulate Matter (PM) – Wild Fires

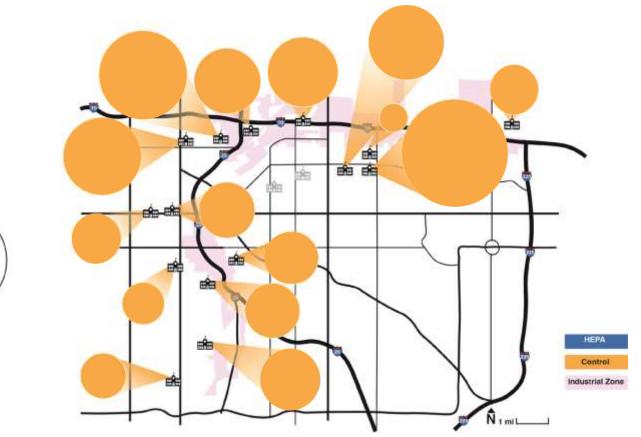
Performance Results

PM2.5 data from the active monitoring system has been used to determine the effectiveness of filtration during the Canadian wildfire events in early June when PM2.5 concentrations were highest during operating hours.

| Location | Outside Air PM2.5 Concentration | Supply Air PM 2.5 Concentration | PM2.5 Removal | Percent Below Recommended Limit |
|-------------------------------------|------------------------------------|------------------------------------|---------------|---------------------------------------|
| Boston Office Building | 60.0 µg/m³ | 6.0 µg/m³ | 90% | 50% |
| New York Office Building | 35.0 µg/m³ | 3.0 µg/m³ | 92% | 75% |
| Washington, D.C. Office Building | 136.0 µg/m³ | 4.2 µg/m³ | 97% | 65% |



Particulate Matter (PM) – Pollution and HEPA Filters



Max [PM ma] = 3 x 10⁴ #/m³

(per occupant)

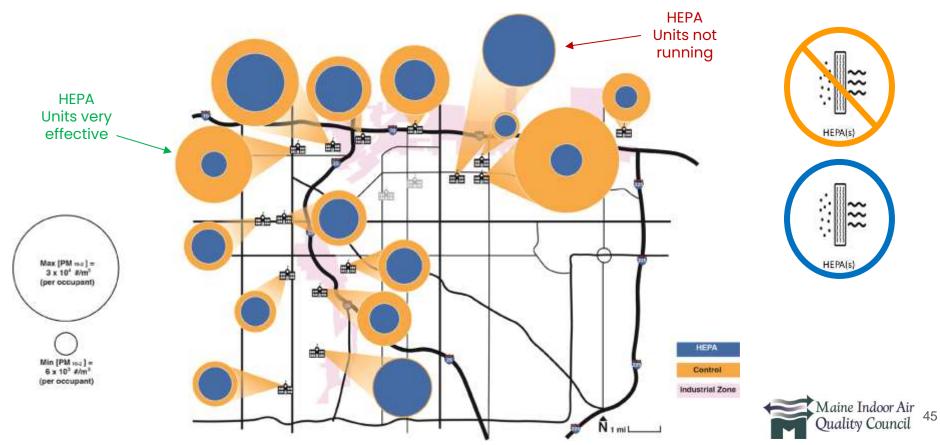
Min [PM use] =

6 x 10³ #/m

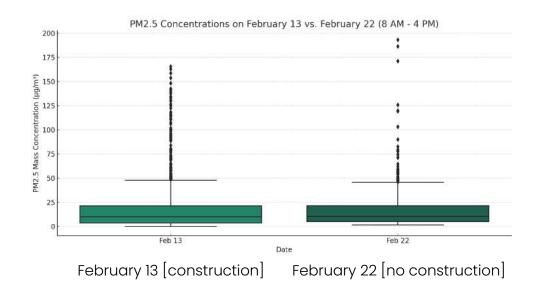




Particulate Matter (PM) – Pollution and HEPA Units



Particulate Matter (PM) – Construction

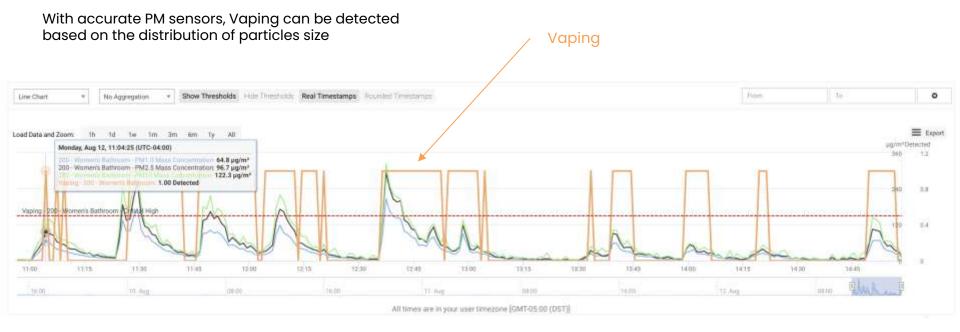


Parents worried construction near a school would impact Air Quality inside the school.

After data analysis, we were able to show that it was not the case, the 2 days are not significantly different

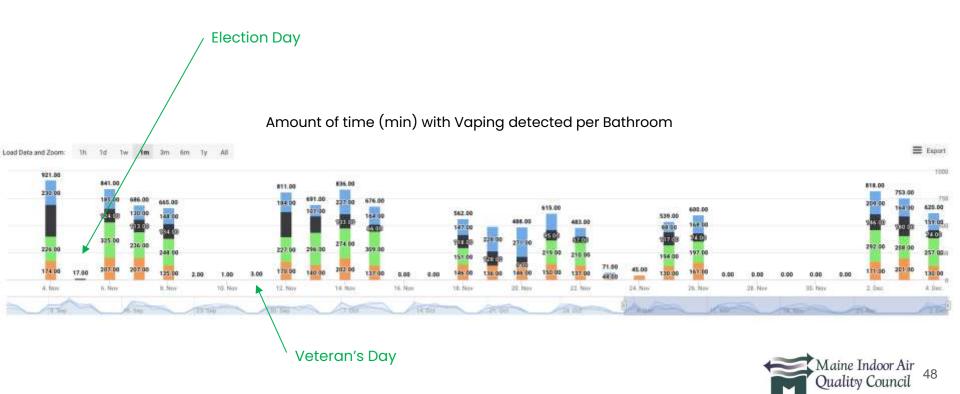


Particulate Matter (PM) – Vaping Detection





Particulate Matter (PM) – Vaping Detection



Activity: Source Identification

Setup

Sensor installed in a large well ventilated space Sources applied right next to the sensor Data collected every 10 seconds

Humidity, CO2, TVOC, and PM2.5 sensors

Possible Answers:

- 1. Vaping
- 2. Candle Burning
- 3. Deodorant Spray
- 4. Ozone Generator
- 5. Mouthwash Breath
- 6. Exhaled Breath
- 7. Febreze Spray
- 8. Coffee Smell





The future of Low Cost Sensors and IAQ Monitoring





Smallest PM Sensor

Machine Learning to detect specific VOCs

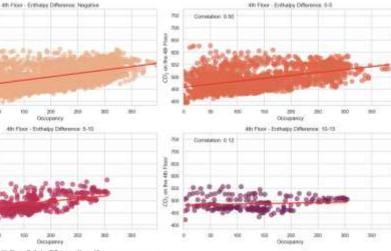


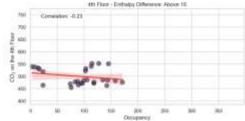
The future of Low Cost Sensors and IAQ Monitoring

350 Correlation 0.58 what is the correlation between occupancy and co2 level for each range of enthalpy 300 difference ? 460 800 200 To assess the correlation between occupancy and CO₂ levels for different ranges of enthalpy difference. 12 800 I will divide the data into bins based on the enthalpy difference and compute the correlation for each range. Let's proceed. 42 Finished analyzing ~ Always expand output? Casy code 764 Committee: 0.48 700 bin_labels + ['very Low', 'Low', 'Moderate', 'High', 'Very High'] 600 sorted_data['Enthalpy Difference Bin'] = pd.cut(sorted_data['Enthalpy Difference'], b \$ 000 correlation results = sorted_data_groupby('Estimates Difference Bin').apply() == bds df correlation_results Enthalpy Difference Bin WORY LOW 0.589797 360 0.373188 700 Moderate 0.489576 600 High 0.282733 1000 Yery High -0,482607 550 stype: float64 8.000

Here's the correlation between occupancy and CO₂ levels on the 11th floor for each range of enthalpy difference

- Very Low Enthalpy Difference: Positive cor ____ tion of 0.59
- Low Enthalpy Difference: Positive correlation of 0.37









Julien Stamatakis - julien@attuneiot.com Attune - info@attuneiot.com