

U.S. EPA STAR Grant

Engage, Educate and Empower California Communities on the Use and Applications of Low-cost Air Monitoring Sensors

During-the-Deployment Workshop



(Conducted December 2018 – April 2019)







Main Objective

Provide communities across California with the knowledge necessary to <u>appropriately</u> select, use, and maintain "low-cost" sensors and to correctly interpret the collected data



Kids Making Sense

Air Sensor Guidebook

Develop new methods to engage, educate, and empower local communities on the use and applications of "low-cost" sensors

Best practices for...

- Sensor deployments
- Data collection
- Data analysis and interpretation
- Next steps: communicating results, planning outreach, developing mitigation strategies





<u>Conduct field and laboratory testing to characterize the performance of commercially-</u>

South Coast

available "low-cost" sensors and to identify candidates for field deployment

- Field Testing:
 - $\circ~$ Sensor tested in triplicates
 - \circ Two months deployment
 - $\circ~$ Comparison with FRM/FEM instruments
 - $_{\odot}~$ Testing performed at a fixed monitoring station









- State-of-the-art characterization chamber
- Particle and gas testing
- $_{\circ}~$ T and RH controlled conditions





Deploy the selected sensors in multiple California communities



and perform a thorough validation and interpretation of the collected data





<u>Communicate the lessons learned to the public and</u> <u>organize outreach activities</u>

Disseminate study results and help answer these key questions:

- ✓ Which tools will be most successful in educating communities to effectively use air monitoring sensors and to engage them in using sensor data?
- ✓ Will a community more likely take action to reduce air pollution exposure when sensors and sensor data are made readily available?
- ✓ Which sensors are the most suitable for community use?
- ✓ How does sensor data quality change with time after sustained use by communities under "real-world" conditions?
- ✓ How do sensor data compare (spatially and temporally) to that of existing monitoring networks?
- ✓ What value is added by these sensors that we are not getting with current network data?



PurpleAir Sensors & Feedback

Example CommunityPM Sensor Network28 sensors installed

Any Feedback on the Following...

- Issues installing or running the sensors?
- Issues accessing/understanding the data?
- Is there any information/training that you feel would help you to make better use of the sensors?
- Are there any questions they would like to know how to answer using the sensors? Or would like to try to answer using sensor data?





Sensor Installation/Non-Installation e-Survey

The Science To Achieve Results (STAR) Grant team at the South Coast AQMD would like to thank you for your participation in the project entitled, "Engage, Educate and Empower California Communities on the Use and Applications of "Low-cost" Air Monitoring Sensors" and to invite you to participate in this very brief online survey about your sensor installation location. Completing this survey with a smart device with a camera will allow you to easily submit a picture.

Installation Survey

Moving forward, please keep an eye out for upcoming community group meetings, an email containing the electronic log note entry form, and changes for end user data visualization and accessibility!



UnsubscribeForward to a friendSouth Coast Air Quality Management District • 21865 Copley Drive, Diamond Bar, CA 91765909-396-2000 • www.aqmd.gov



Log-book e-Survey – Adding Context to the Data

(also available

as a hard copy)



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Any observations by the community?



Sensor Data

What can we do with this data?

- Assess spatial and temporal variability
- Compare regional and local trends across network
- Evaluate impact of wind speed & wind direction
- Determine when particle or gas pollution is high/low
- Identify potential nearby pollution sources











Example Analysis – Complete Sensor Data



- All sensors plotted together
- Darker indicates overlapping, lighter indicates a single sensor signal



Example Analysis – Temporal Trends

- On average, higher PM at night and lower during the day
 - Expected trend driven by daily fluctuations of the "planetary boundary layer"
- Also, the averages for each month are similar, with the exception of August where it is much higher average





1000

800

000

400

200

0

08/18

09/18

⊃М_{2.5} (µg m⁻³)

Example Analysis – Impacts from a Nearby Wildfire

• Shortly after the fire began we see large enhancements impacting the all of the local sensors





Example Analysis – Impacts from a Nearby Wildfire

- Average for each site from 8/8/18 – 8/15/18
- Average for all sites is relatively high
- There is also a clear gradient with some sites seeing higher peaks than others





Example Analysis – Comparisons

- By comparison, the averages for other months are much lower (Aug: $15 76 \mu g m^{-3}$)
- The spatial trends also differ from what was seen during the fire







Example – Studying Potential Local Sources

- In the fall, there are repeated short-term enhancements at one site
- This may be an indication of a local source impacting a sensor – though it would be important to verify that these "events" are not an issue with the sensor







Example – Studying Potential Local Sources

 Assuming these are events driven by a source, supplemental data (e.g., wind speed and direction) can help us to learn more about the location of the potential source







Example – Studying Potential Local Sources

- Data from another site at which there were also fairly substantial short-term enhancements
- An enhancement like this may be due to local activities, such as cooking or grilling







Summarizing the Data

- Daily average throughout the year
- Short-term events may have a slight impact on averages
- Wildfire impacts from August still stand out as a major event



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One Final Example

- Data: May 7th May 9th 2019
- Sensors seem well suited to provide real-time, localized information

NEWS > CRIME + PUBLIC SAFETY

Chino dairy hay fire continues to burn; warning about smoke issued

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Conclusions

- The low-cost sensors reflect expected air quality trends
- These sensors reveal interesting trends across the network and help to track major events
- It seems there is also potential to use these sensors to learn more about local sources



Any Final Thoughts?

(1) What result that was presented did you find most interesting or unexpected? Why?

(2) What result was the most actionable, or the most relevant to your concerns? Would you change your behavior based on this results or did it give you any ideas of how we may be able to improve air quality using sensors?

(3) Was there a particular plot, or visual, or story regarding the data that you found most easy to connect with? In other words, what way of presenting data do find most effective?



Next Steps

<u>Develop a cloud-based computing platform</u> <u>to ingest, store, analyze, and display sensor data</u>

Data analysis workloads larger than typical tools can handle Fence-line monitoring: ~15 million rows of data Regional monitoring network: ~40 million rows of data STAR Grant: ~50 million rows of data South Coast AQMD R1180: XX million rows of data CA AB617: X billion rows of data



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Thank you!