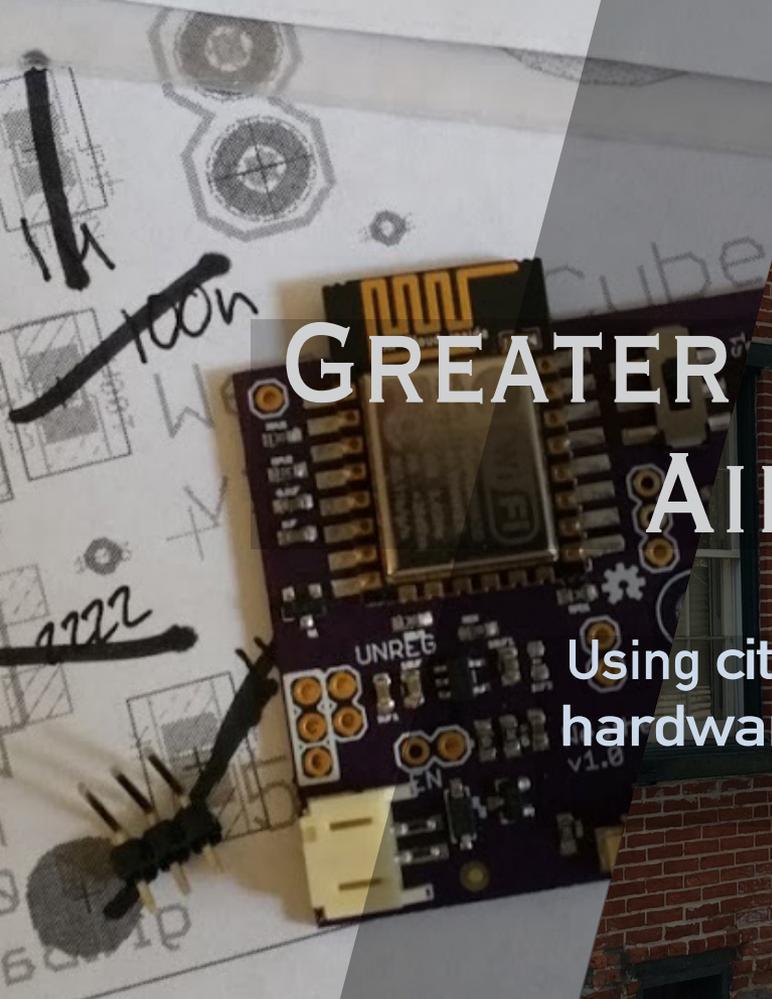


License No.: 044020213



GREATER BALTIMORE OPEN AIR PROJECT

Using citizen science, IoT, & open hardware to measure air quality in Baltimore



Baltimore Open Air Partners



... and many others



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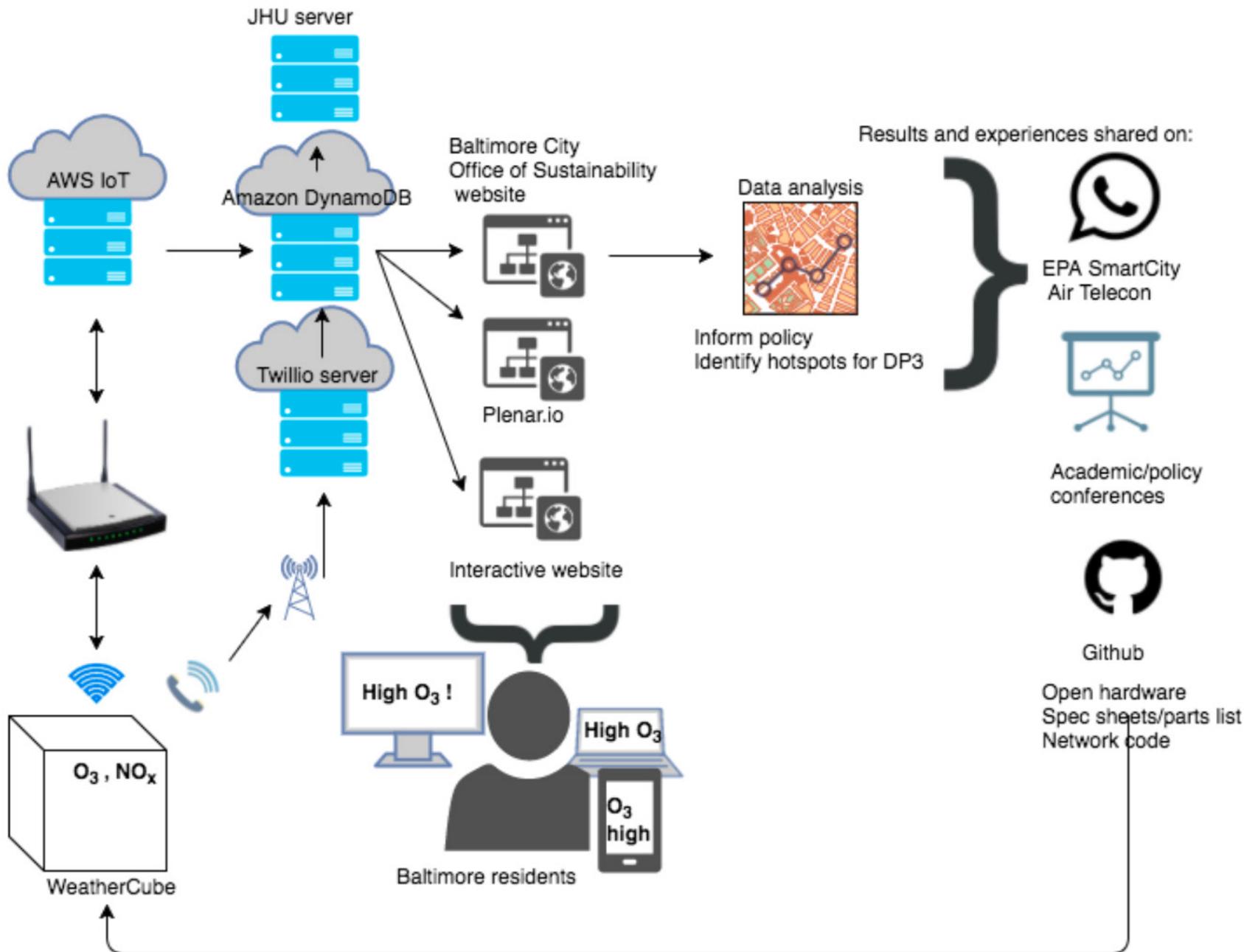


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Specifically, we aim to:

Design and build 300 air quality monitors, working with **citizen scientists** to build and deploy the network, as well as manage the **data & quantify** spatial variability in **air quality**.

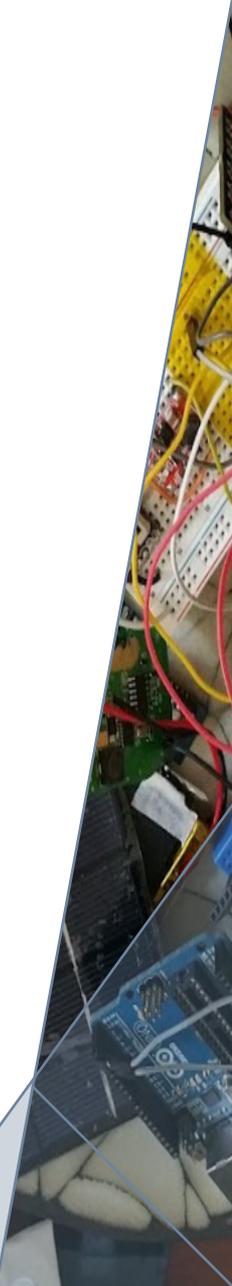
OUR AP- PROACH

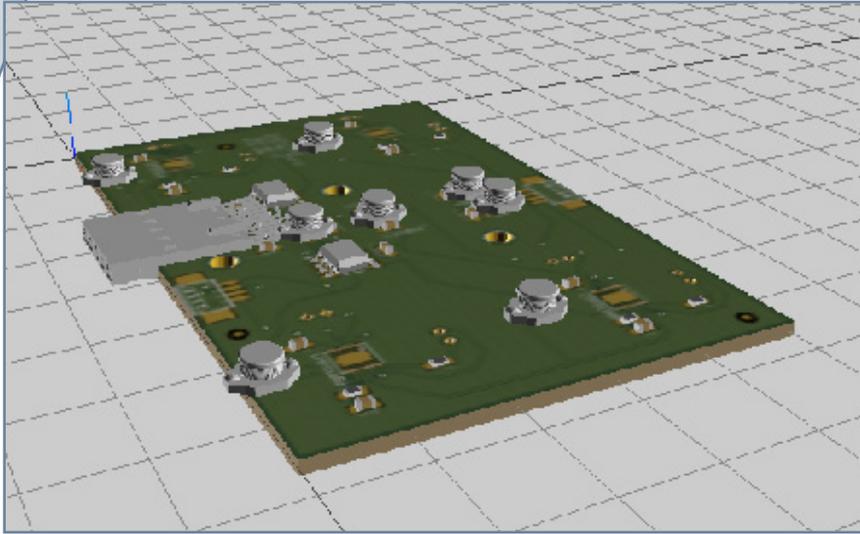


HARD-
WARE

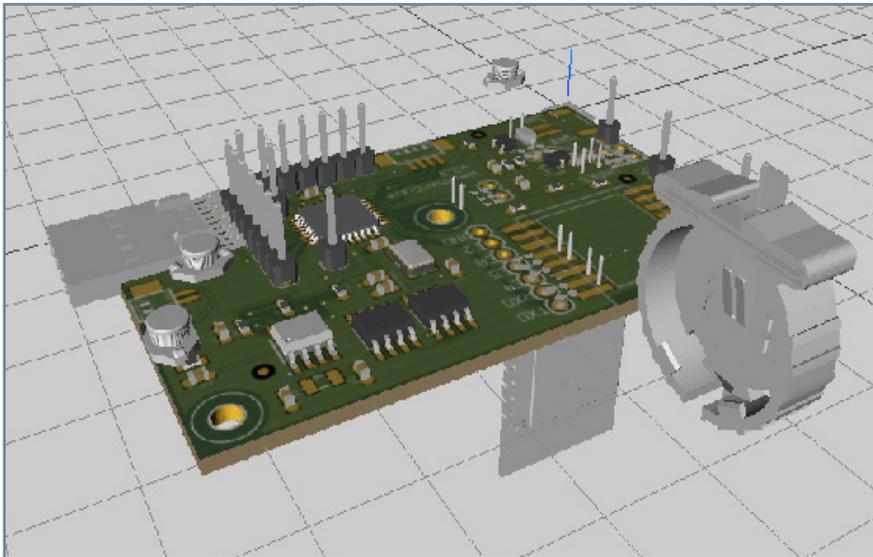
Circuit Design

Chris Kelly is a PhD student in the Department of Geography and Environmental Engineering at JHU. His research interests include software and hardware development for water and sanitation monitoring, multi-scale spatial estimates of water and sanitation treatment access, the use of environmental and epidemiological data in infrastructure planning, collaborative GIS, and collaborative infrastructure funding mechanisms.

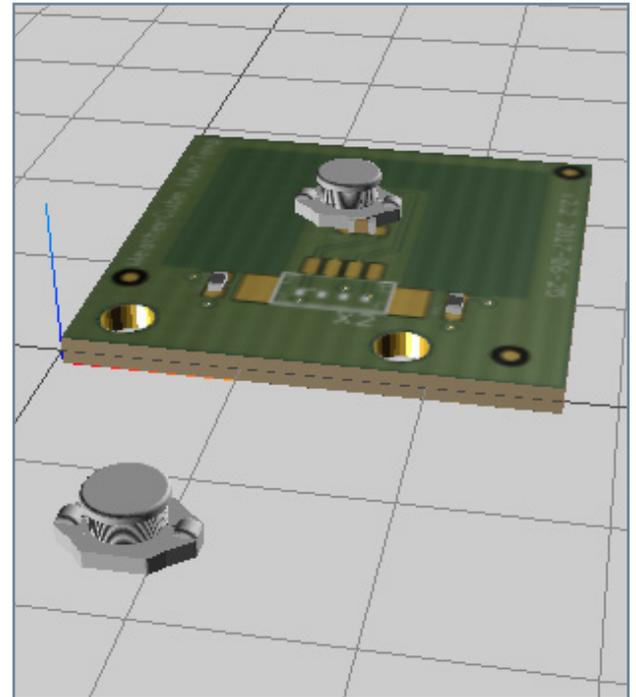




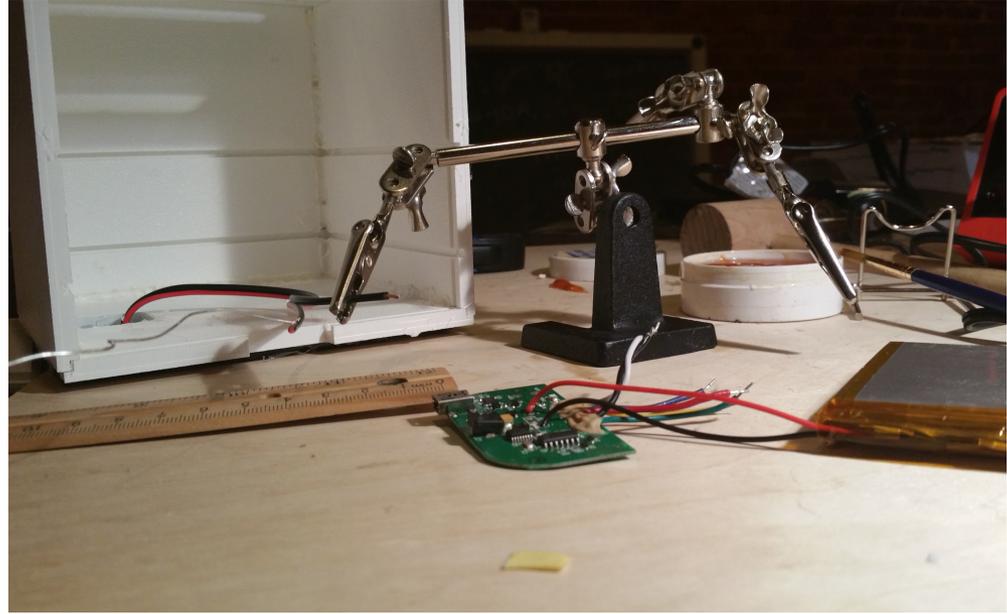
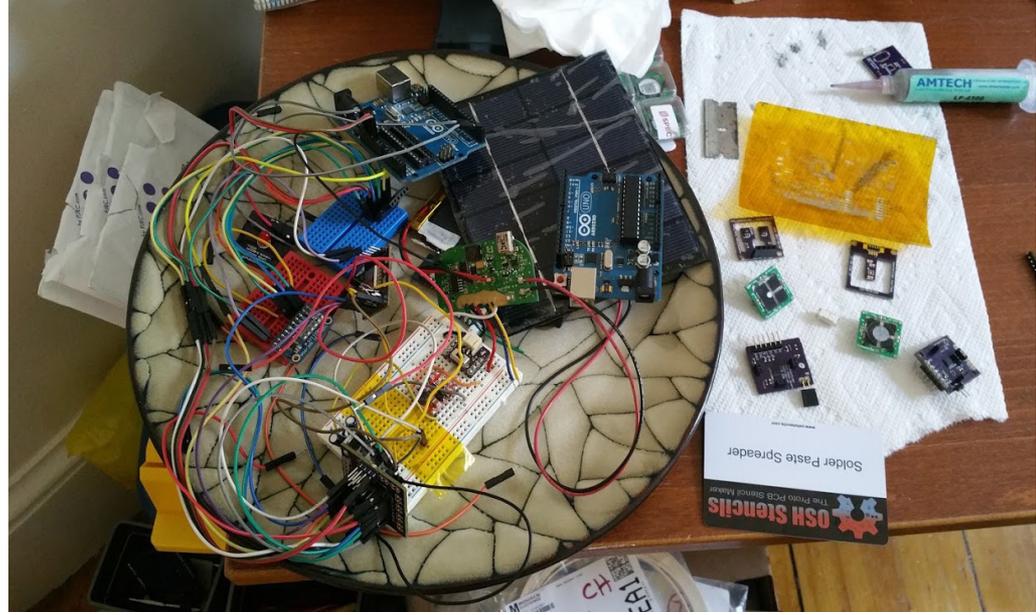
LMP9100



Main Board



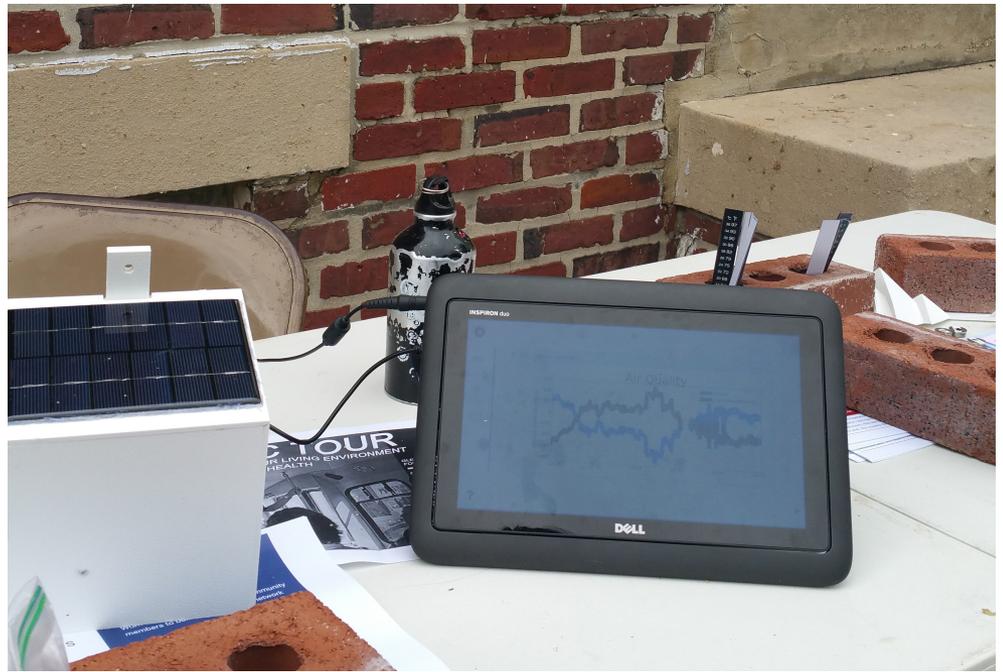
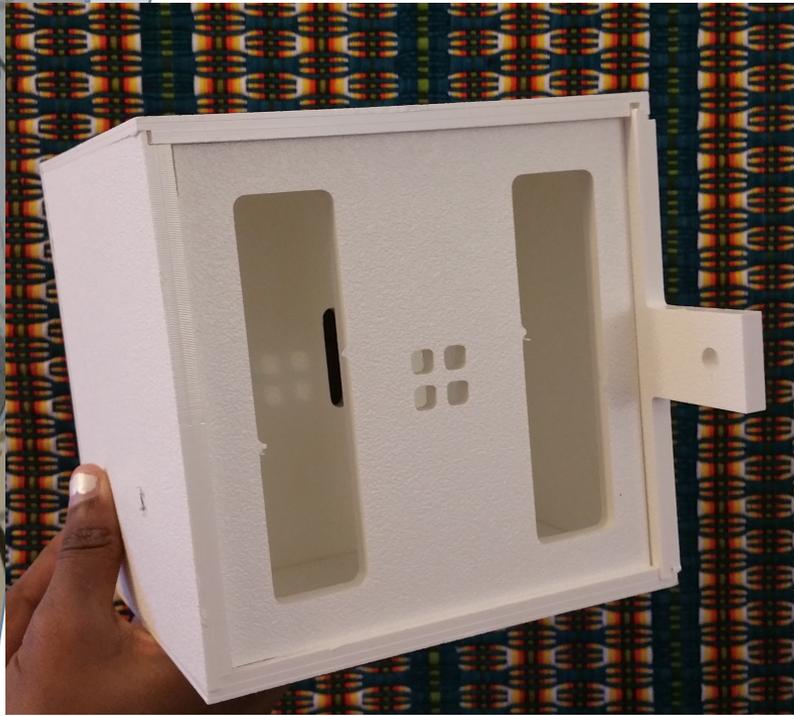
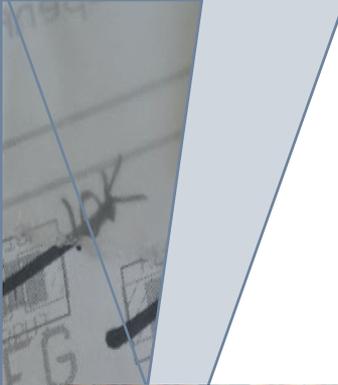
Humidity



Case Design

Yan Azdoud is a Postdoctoral fellow in the civil engineering department at Johns Hopkins University. He is interested in computational modeling in solid mechanics and has broader interest in interdisciplinary research.







Upcoming hardware improvements

- » Cellular modem integrated into circuit board (expected Aug. 2017)
- » Wired connectivity (Aug. 2017)
- » More compact circuit board (Nov. 2017)

SOFT-
WARE

Browser tabs: Inbox (16,477) - gottscott@... x Greater Baltimore Open Air B... x Baltimore Open Air Air Sens... x DynamoDB - AWS Console x

Address bar: Secure https://us-west-2.console.aws.amazon.com/dynamodb/home?region=us-west-2#tables:selected=ESP8266AWSDemo

Navigation: Services Resource Groups AWS IoT DynamoDB

User: Anne A Scott Oregon Support

DynamoDB

Dashboard

Tables

Reserved capacity

DAX

Dashboard

Clusters

Subnet groups

Parameter groups

Events

[Create table](#) [Actions](#)

Filter by table name

Name
<input checked="" type="radio"/> ESP8266AWSDemo
<input type="radio"/> LambdaTesting

ESP8266AWSDemo

[Close](#)

[Overview](#) [Items](#) [Metrics](#) [Alarms](#) [Capacity](#) [Indexes](#) [Triggers](#) [Access control](#) [Tags](#)

[Create item](#) [Actions](#)

Scan: [Table] ESP8266AWSDemo: id, timest

Viewing 1 to 55 items

Scan: [Table] ESP8266AWSDemo: id, timest

Add filter

Start search

<input type="checkbox"/>	id	timest	temp
<input type="checkbox"/>	ESP01	20000101000...	21
<input type="checkbox"/>	ESP01	20000101000...	27
<input type="checkbox"/>	ESP01	20000101000...	24
<input type="checkbox"/>	ESP01	20000101000...	23
<input type="checkbox"/>	ESP01	20000101000...	26
<input type="checkbox"/>	ESP01	20170612032...	20
<input type="checkbox"/>	ESP01	20170612032...	29
<input type="checkbox"/>	ESP01	20170624163...	26
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<input type="checkbox"/>	ESP01	20170624163...	22
<input type="checkbox"/>	ESP01	20170624163...	22



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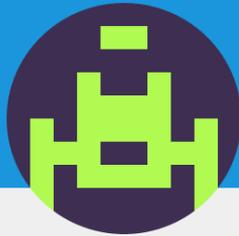
External Resources

- ckan
HYDROSHARE
Dataset Storage
- Windows Azure
amazon web services
Cloud Computing

Tethys Software Suite

- HTCondor
Distributed Computing
- GeoServer
Spatial Publishing
- s2north
Geoprocessing
- PostGIS
Spatial Database
- Web GIS
- Google maps
OpenLayers™
Highcharts JS
Visualization

Tethys Software Suite



admin

Name

First Name:

Last Name:

Credentials

Username: admin

Password: [Change Password](#)

API Key

Token:

Gravatar

Email:

Account

Member Since: June 27, 2017

Status: Active

Delete Account: [Delete Account](#)

Dam Inventory

Navigation

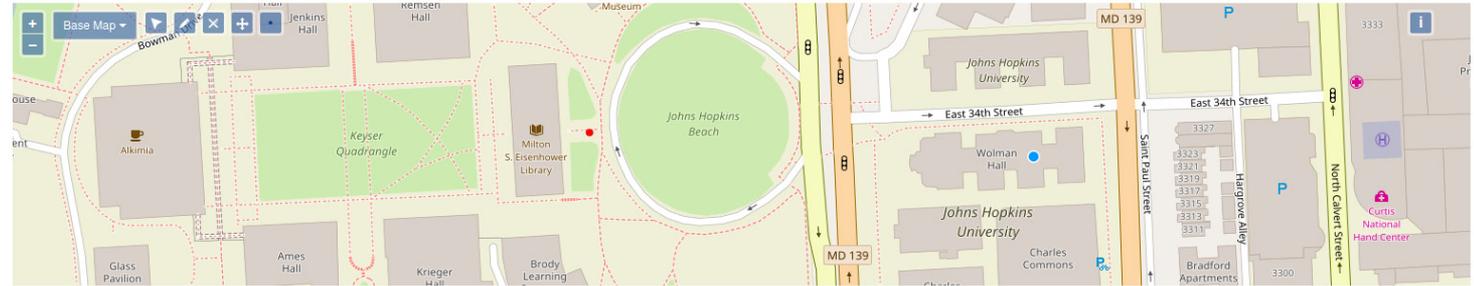
Home

Sensors

Add Sensor

Add Sensor

Location



ID

BLT01

Sensor Type

Temperature

Comment

Between the Milton S Eisenhower Library and the Beach

Date Installed

June 17, 2014

+ Add

Cancel

Dam Inventory

Navigation

[Home](#)

Sensors

[Add Sensor](#)

Air Quality Sensors

Show entries

ID	Type	Comments	Date Installed
BLC01	CO2	Alternate Sensor Model	July 21, 2016
BLT01	Temperature	Between the Milton S Eisenhower Library and the Beach	June 17, 2014
BLT02	Temperature	Canton Waterfront intersection	July 16, 2015

Showing 1 to 3 of 3 entries

Previous Next

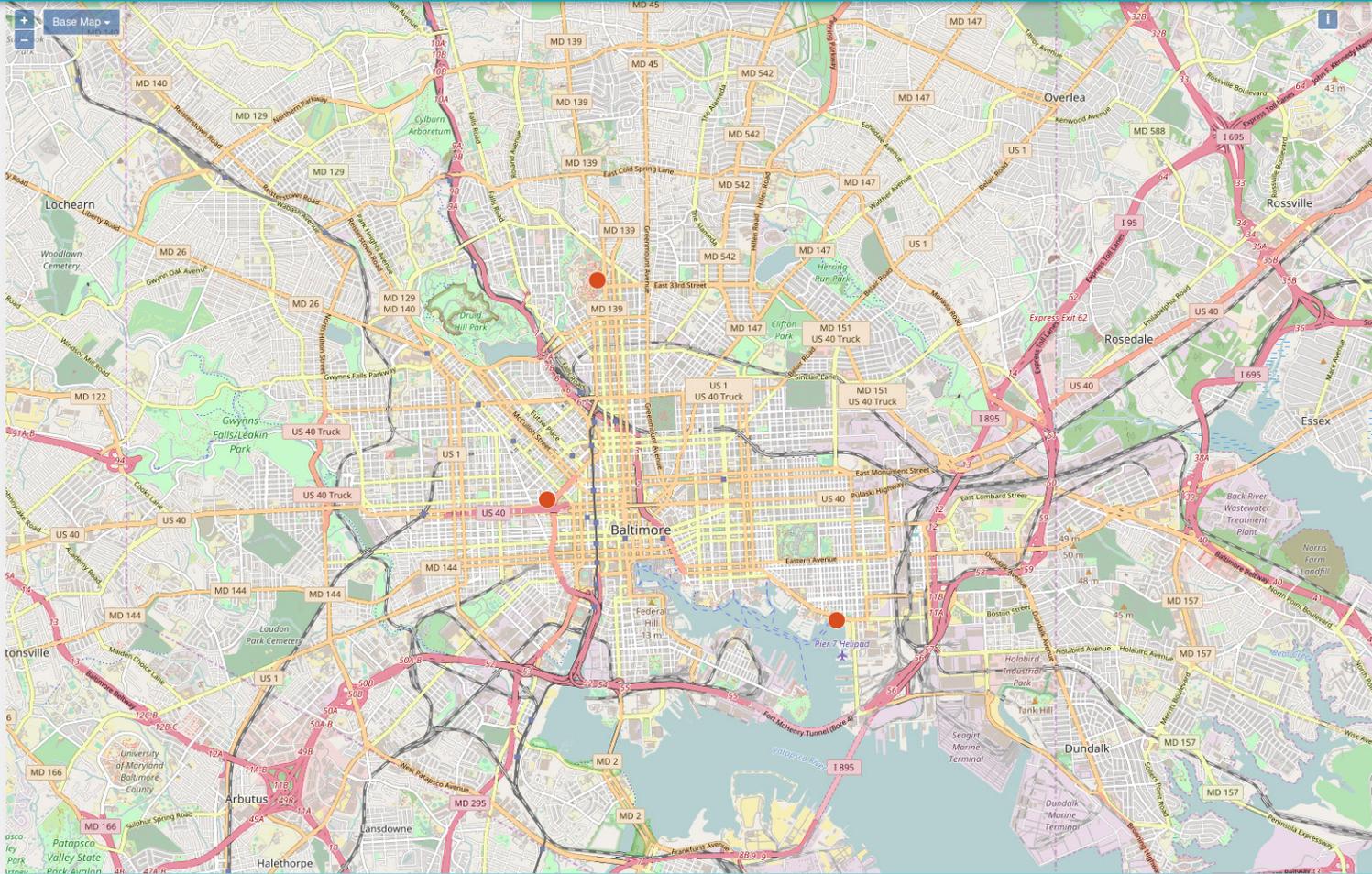
Dam Inventory

Navigation

Home

Sensors

Add Sensor



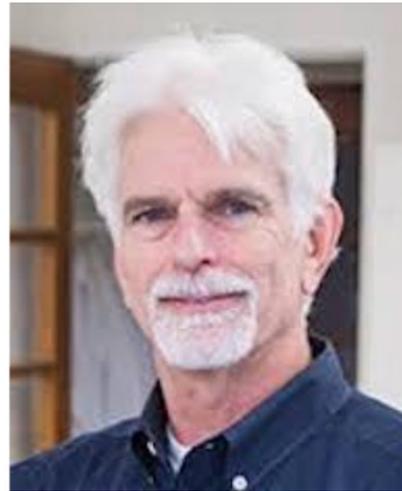
+ Add Sensor

This spring, we have:

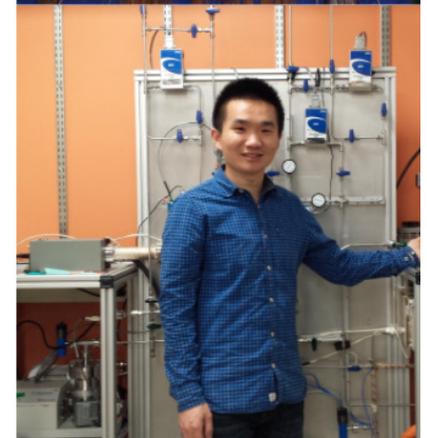
- » Designed a wireless, solar powered air quality monitoring platform
- » Began laboratory calibrations of air quality sensors
- » Created computer code to upload data to Amazon Web Services (AWS) database DynamoDB
- » Began writing code to utilize AWS IoT lambda rule engine
- » Hosted community workshops
- » Began coordinating with site hosts for monitor installation
- » Continued outreach online: website, twitter

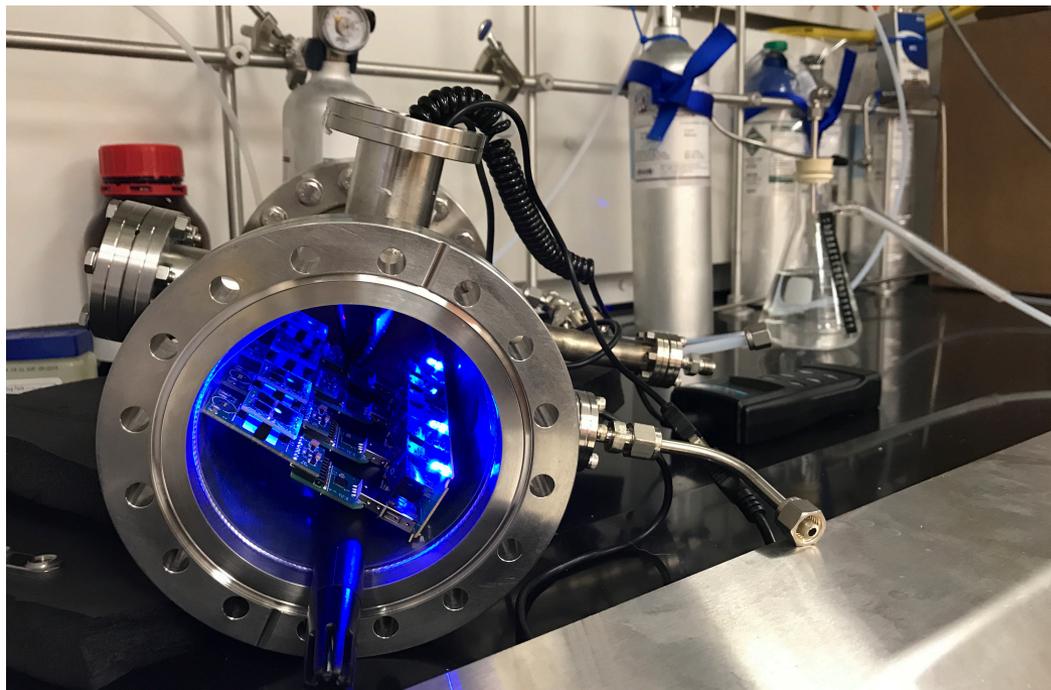
CALIBRA- TION

Calibrations took place at University of Maryland and in Professor Sarah Horst's lab at Johns Hopkins University.



UNIVERSITY OF
MARYLAND





Handwritten notes: "I made mistakes and forgot to write I could" and "I used the same method: letting gasses mix until the same graph readings reach a solid asymptote"

O ₂ set value (%)	O ₂ read value (%)	Set O ₂ (ppm)	Desired O ₂ value	Flow rate (L/min)	RH + temp (°C)	Time (min)	Flow rate (L/min)	Final (ppm)
150	146.7-152	0.01	92.00	92.00	145.8°C	75.17	25.4°C	31.30.94
0	0-2	0.01	92.08	92.00	0.00	73.9°C	23.5°C	31.15.0
100	99-104	0.005	92.02	98.49	98.49	76.57	23.4°C	30.00.0
10	49-50	0.045	46.64	37.31	39.47	77.17	21.4°C	30.00.0
20	19-20	0.06	14.74	31.52	14.94	77.07	21.4°C	33.00.0
40	9.1-10	0.001	10.43	19.85	39.77	77.07	21.4°C	33.00.0

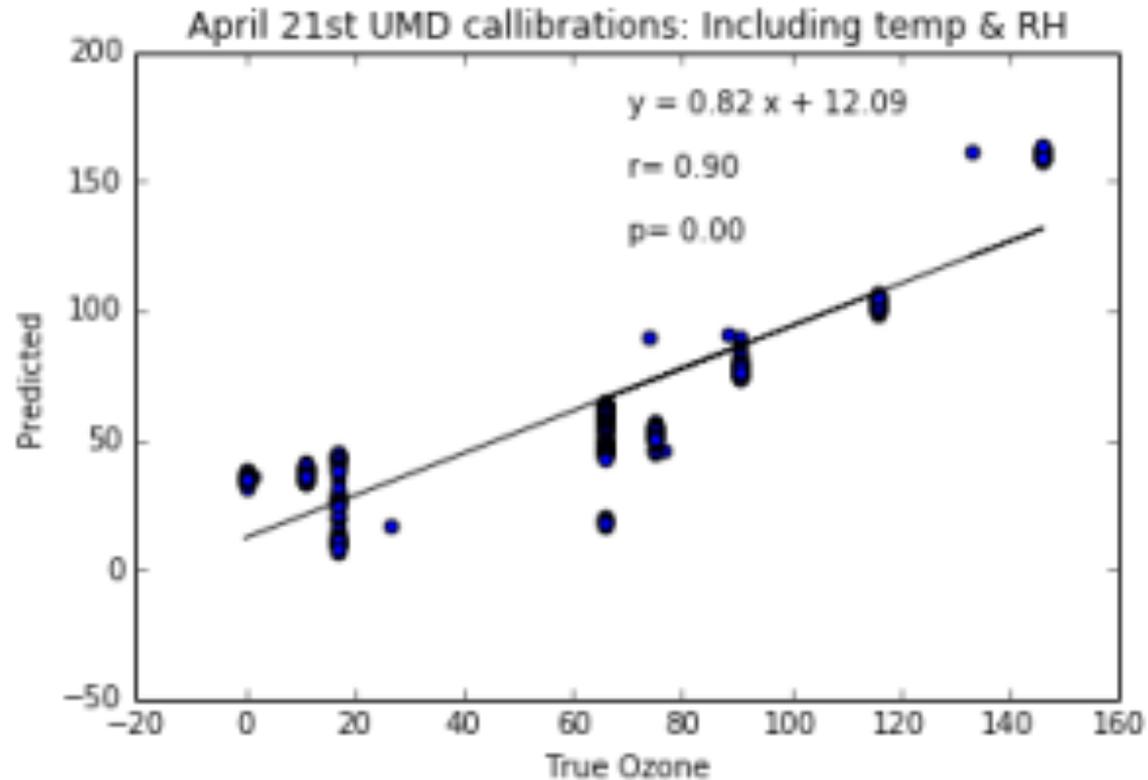
Handwritten notes: "I used the same method: letting gasses mix until the same graph readings reach a solid asymptote"

O ₂ set value (%)	O ₂ read value (%)	Set O ₂ (ppm)	Desired O ₂ value	Flow rate (L/min)	RH + temp (°C)	Time (min)	Flow rate (L/min)	Final (ppm)
150	150.0	0.01	147.21	77.37	77.37	15:00.00	23.7°C	31.30.94
0	0-1.5	0	0	77.07	77.07	15:00.00	23.2°C	31.15.0
100	100.2	0.005	93.80	99.06	99.06	15:00.00	23.2°C	30.00.0
10	9.1-10	0.001	37.73	39.84	39.84	15:00.00	23.2°C	30.00.0
20	19.5	0.001	14.93	19.96	19.96	15:00.00	23.2°C	30.00.0
40	9.1-10	0	40	77.27	77.27	15:00.00	23.3°C	30.00.0

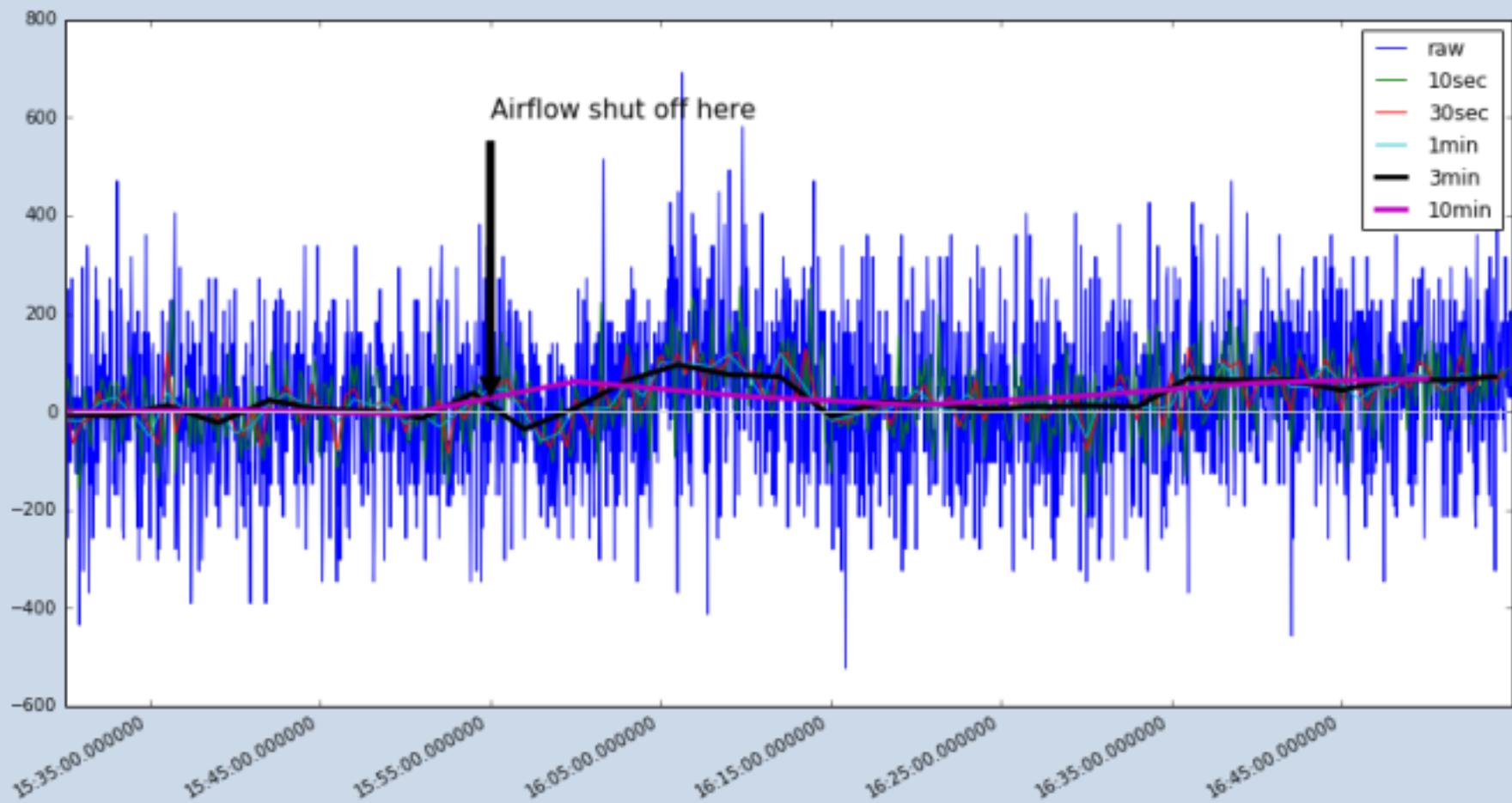
12-Jun	O3 Set Value (duration (min))	O3 Read Value	total flow rate	RH (%)	Start time
	150 30	150.7	1.924	81.2	Didn't record
	0 15	21	1.953	80.1	14:20
	100 15	99.6	1.933	79.9	14:43
	25 15	24.2	1.953	79.3	15:20
	10 15	10	1.933	78.6	15:43
	0 15	2	1.936	77	16:03

lab_df['o3'].loc['2017-06-12 14:20:01':'2017-06-12 14:43:00'] = 21
lab_df['o3'].loc['2017-06-12 14:43:01':'2017-06-12 15:20:00'] = 100
lab_df['o3'].loc['2017-06-12 15:20:00':'2017-06-12 15:43:00'] = 24
lab_df['o3'].loc['2017-06-12 15:43:00':'2017-06-12 16:03:00'] = 10
lab_df['o3'].loc['2017-06-12 16:03:00':'2017-06-12 16:18:00'] = 2

Sensor Calibrations

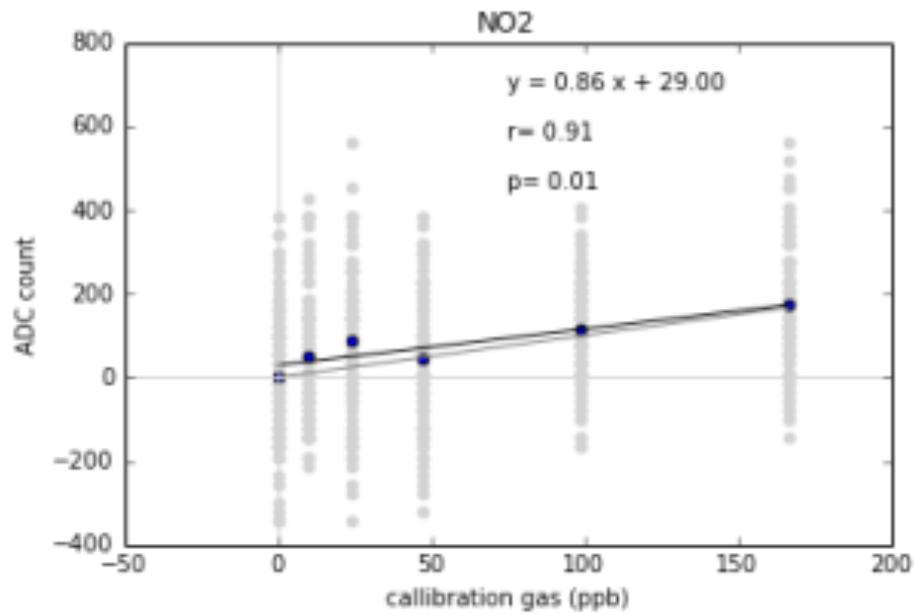
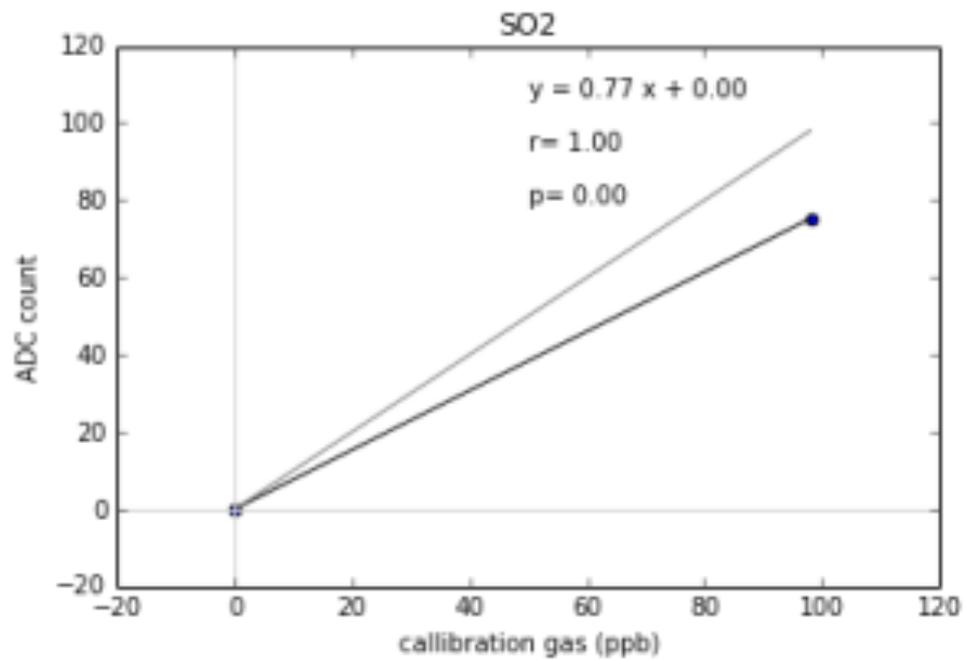
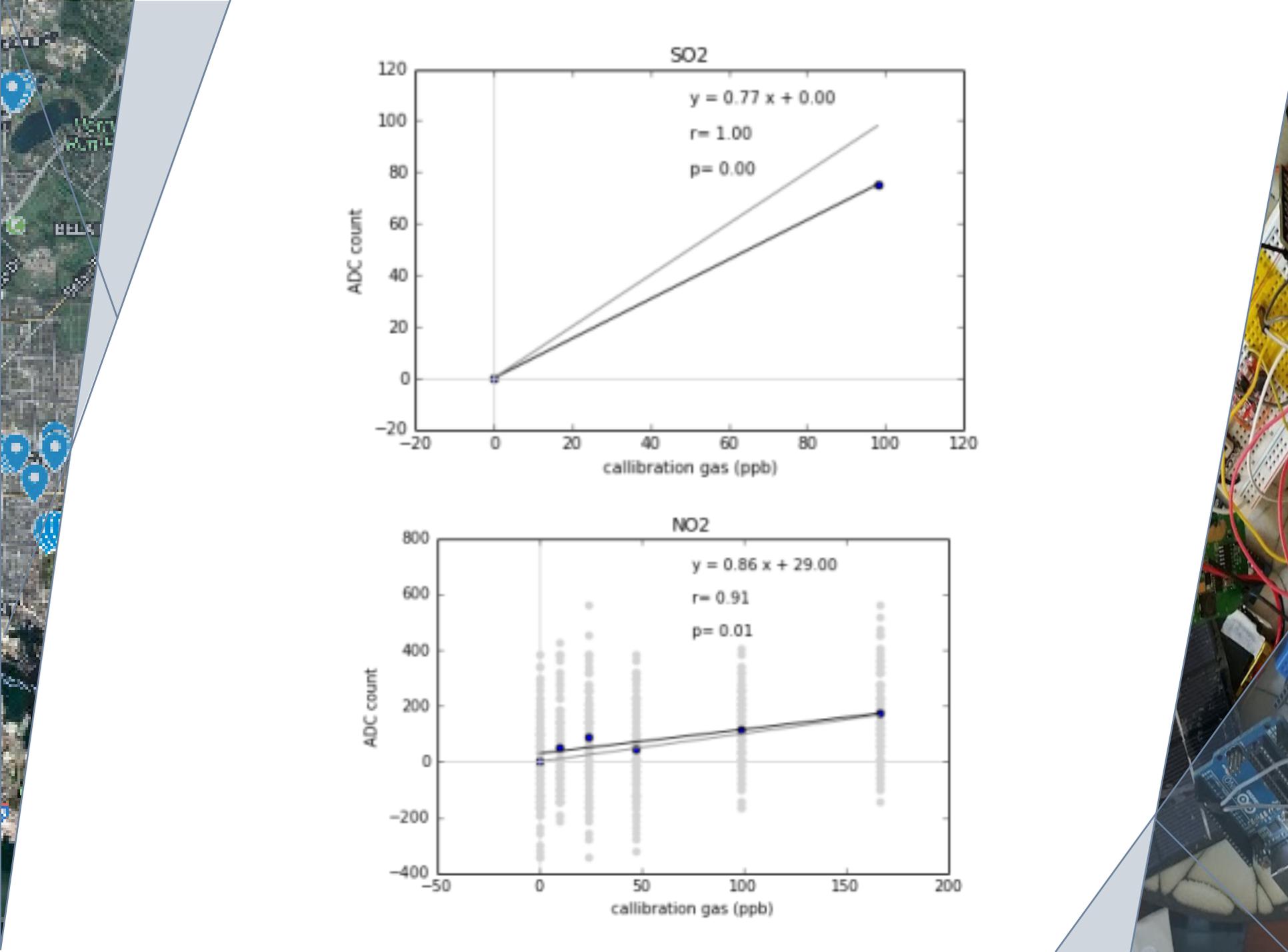


Protocol following: McElroy, F. F. "Transfer standards for calibration of air monitoring analyzers for ozone." EPA Technical Assistance Document (1979).



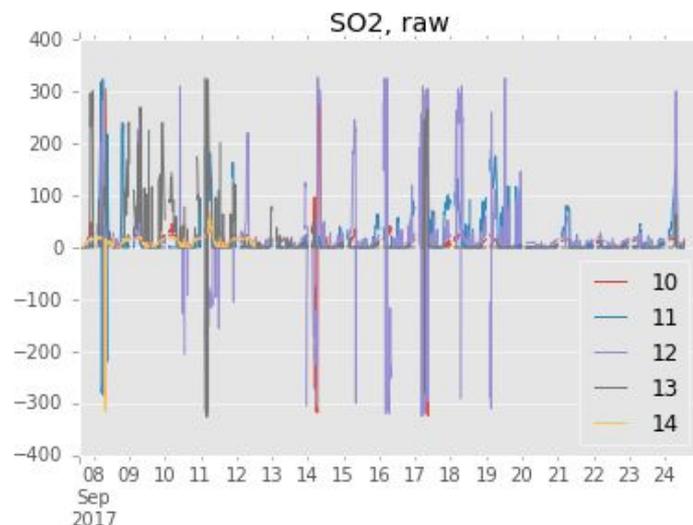
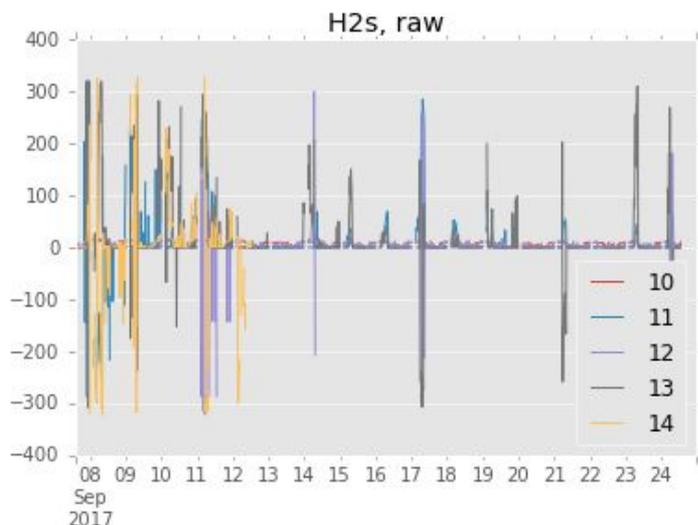
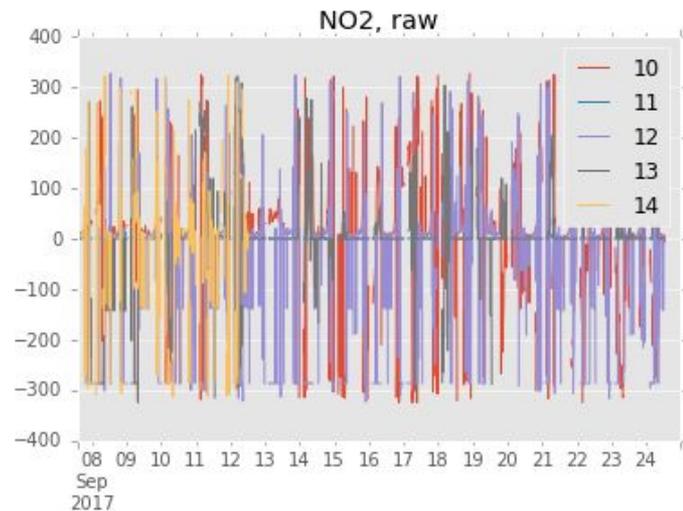
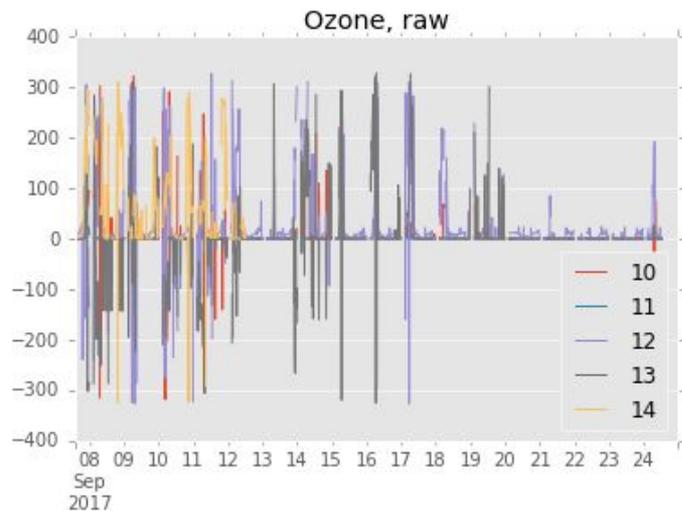
Calibration of NO2 sensor at 0ppb





Raw data

Raw ADC counts are very noisy.



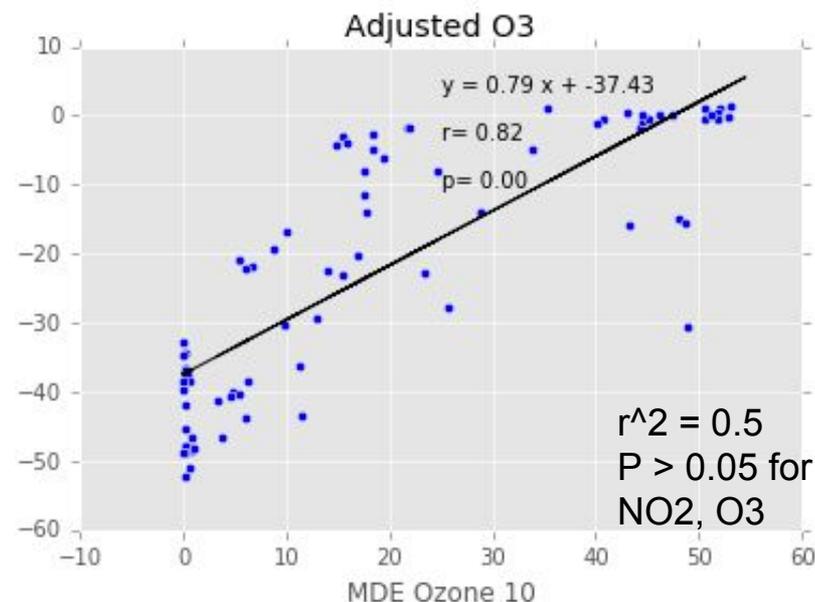
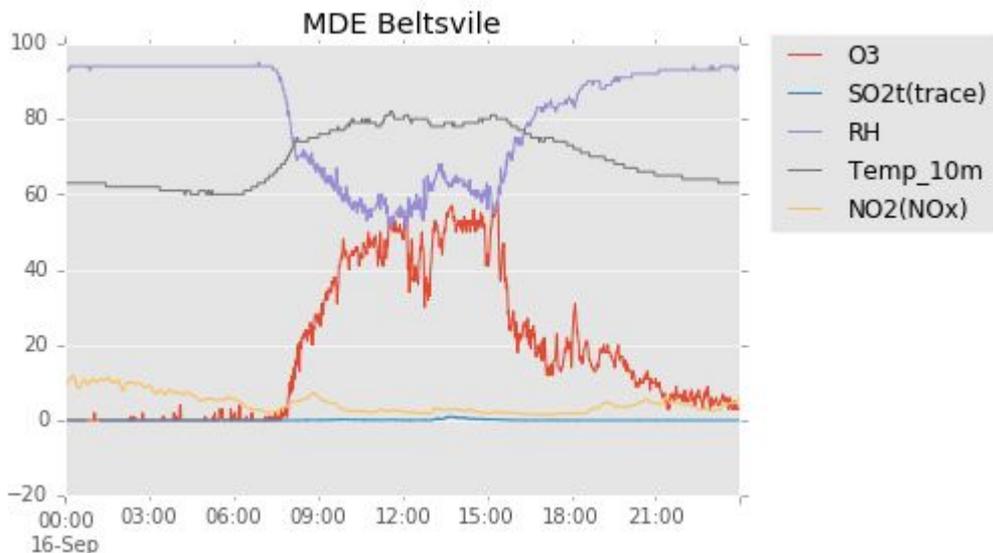
Sensor calibrations

$$[NO_2]_{sensor} = x_1[NO_2]_{true} + x_2[O_3] + x_3[H_2S] + x_4[SO_2] + \dots$$

Sensor values are a function of the target gas, but also other gasses. We want to solve for:

$$[NO_2]_{true} = 1/x_1 \cdot (-[NO_2]_{sensor} + x_2[O_3] + x_3[H_2S] + x_4[SO_2] + \dots)$$

Ozone sensor calibration



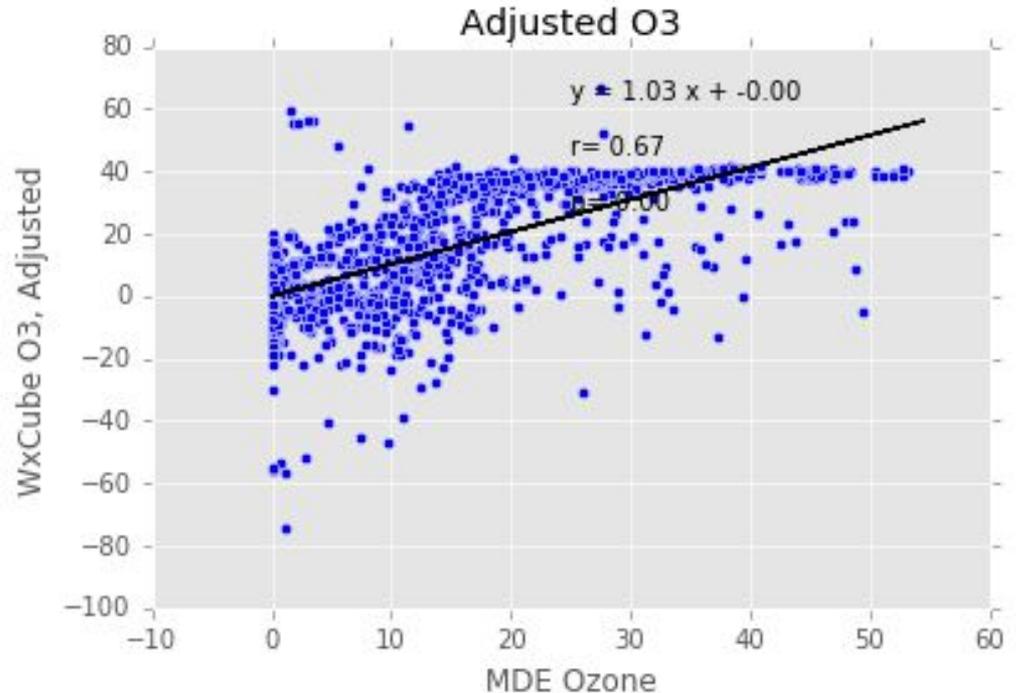
$$O_3 = -0.55 + -0.07 \cdot [O_3]_{true} \\ + 0.036 \cdot [NO_2] + 0.187 \cdot [SO_2] + 3.62 \cdot [H_2S]$$

Ozone sensor calibration

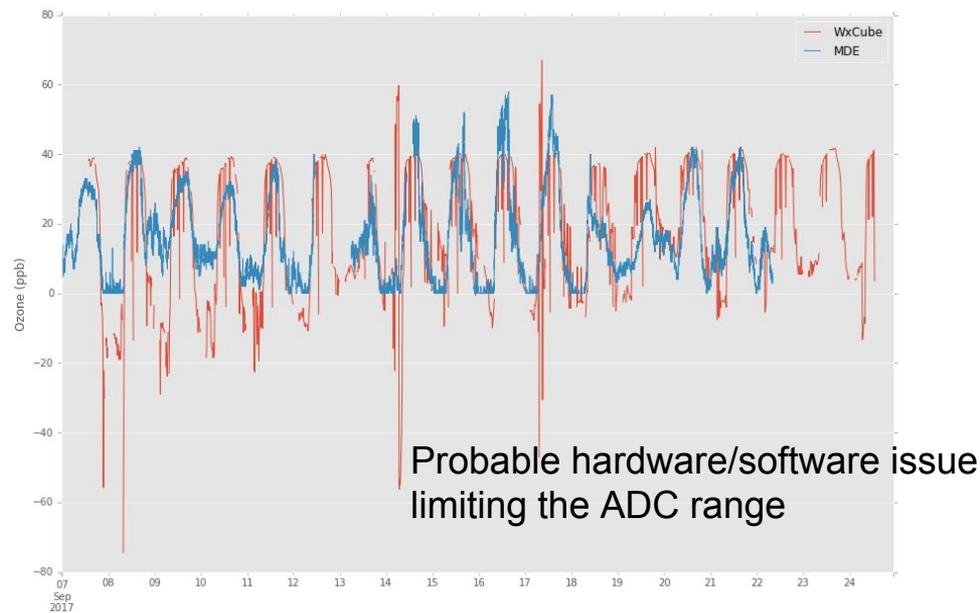
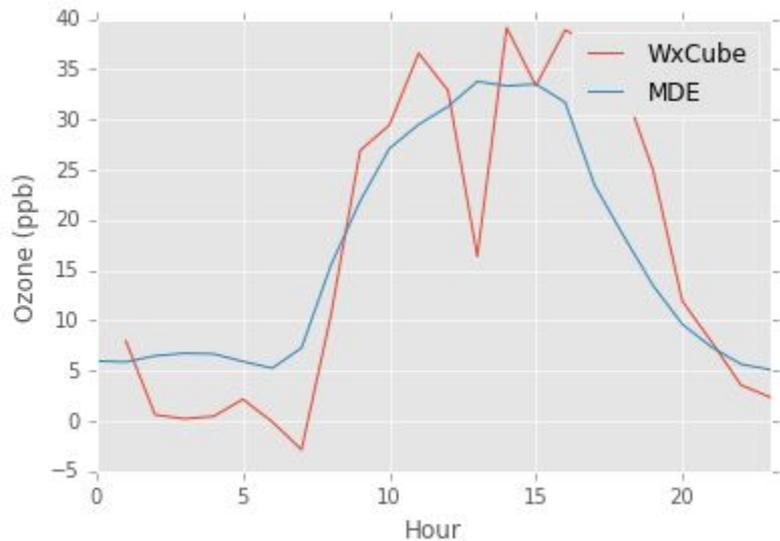
The previous result assumes that the true ozone is known. Here, we remove known ozone and apply the regression coefficients:

$$O_3 = -0.55 + 0.036 \cdot [NO_2] + 0.187 \cdot [SO_2] + 3.62 \cdot [H_2S]$$

Mean error = 1.05 ppb.

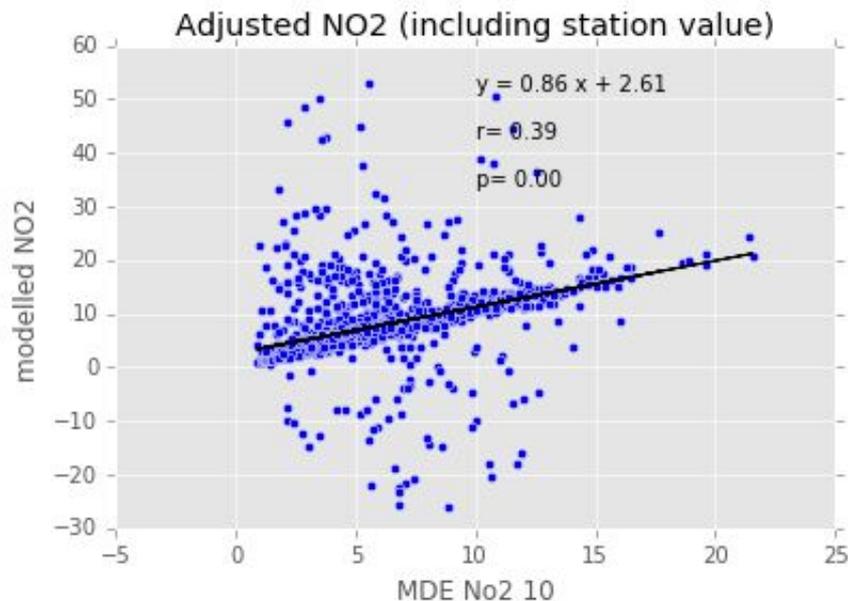
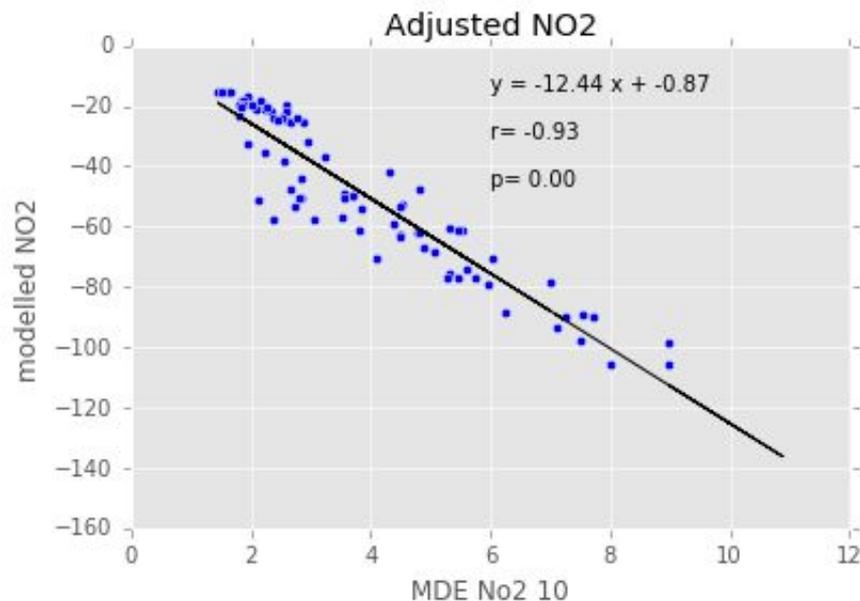


Ozone sensor calibration



Final regression captures the diurnal cycle of ozone well, but not the range

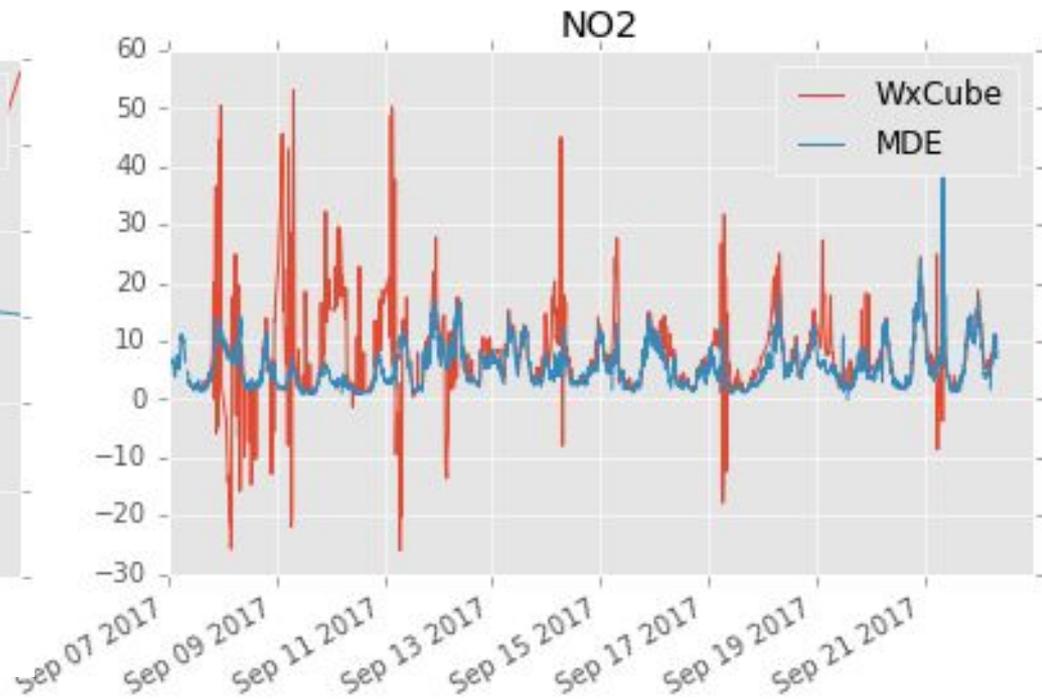
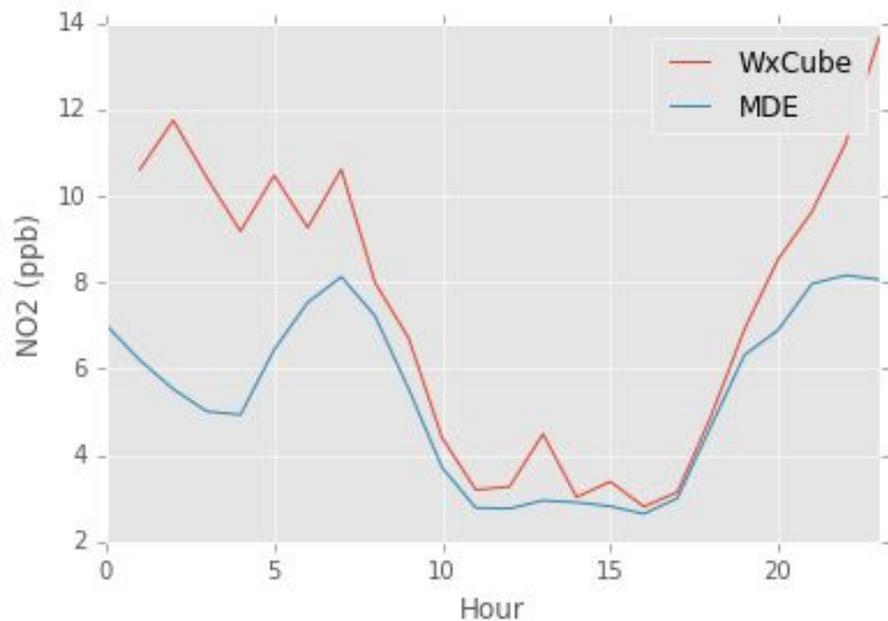
NO2 sensor calibration



$$NO_2 = 20.8 - 8.13 \cdot [NO_2]_{true} - 0.29 \cdot [SO_2] - 2.58 \cdot [H_2S] + 0.1 \cdot [O_3]$$

$r^2 = 0.385$
 $p > 0.05$ for O_3

NO2 sensor calibration



Mean difference = -1.91 ppb

Sensor calibrations

We used a day of data for one sensor and compared it to Maryland Department of the Environment (MDE) data.

Sensors were co-located.

Our sensors report 3 minute averages every 12 minutes.

MDE samples at 1 minute frequency (data has not undergone QA procedures).

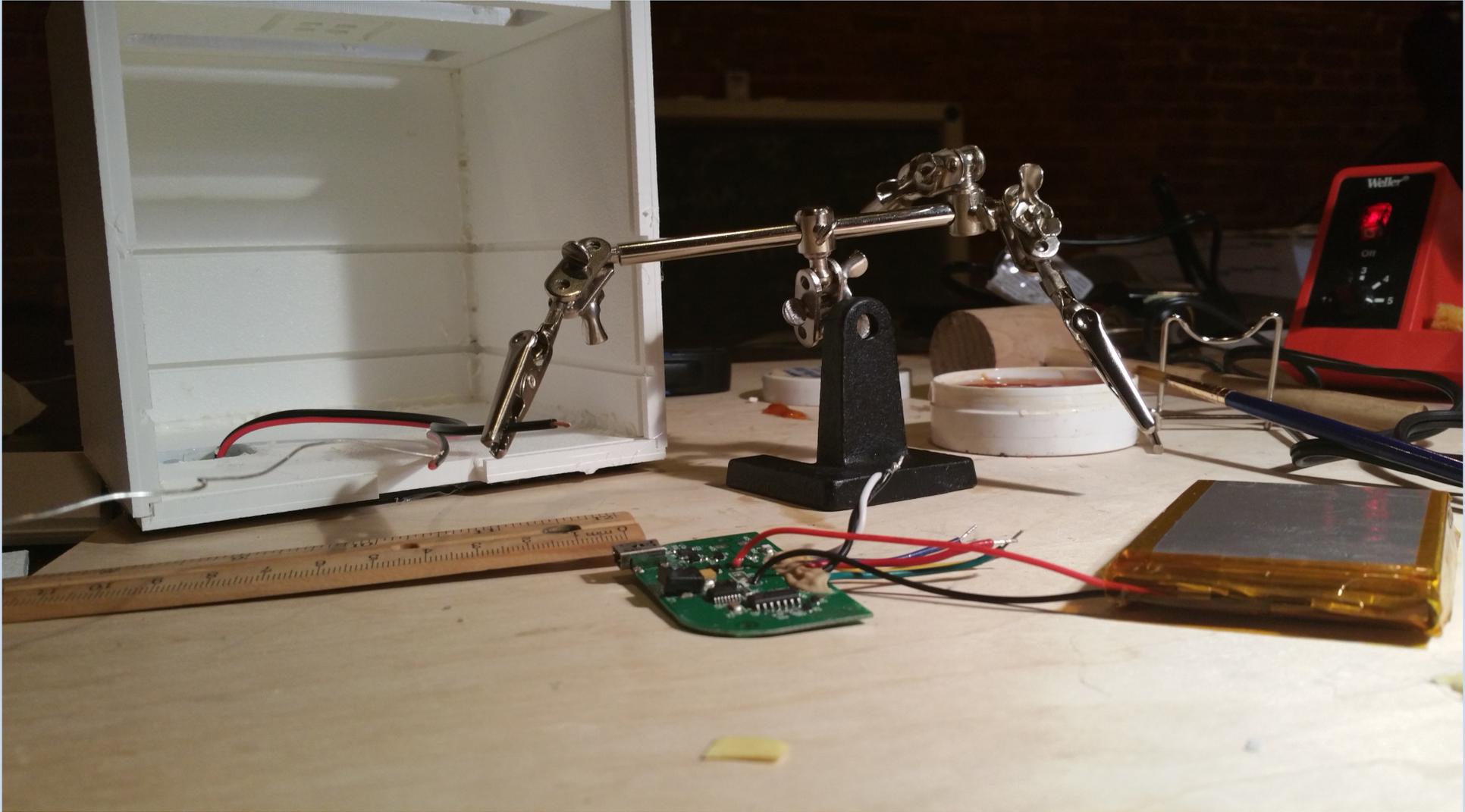
A linear regression was computed in Python (statsmodels ols function).



ASSEM-
BLY









Upcoming challenges:

Manufacturing July 19-

Field calibration August X-

Deployment with community centers in August

Visualization website in late July

bmoreopenair@gmail.com

ascott47@jhu.edu

@bmoreopenair