

Remotely calibrating gas sensor devices in the field

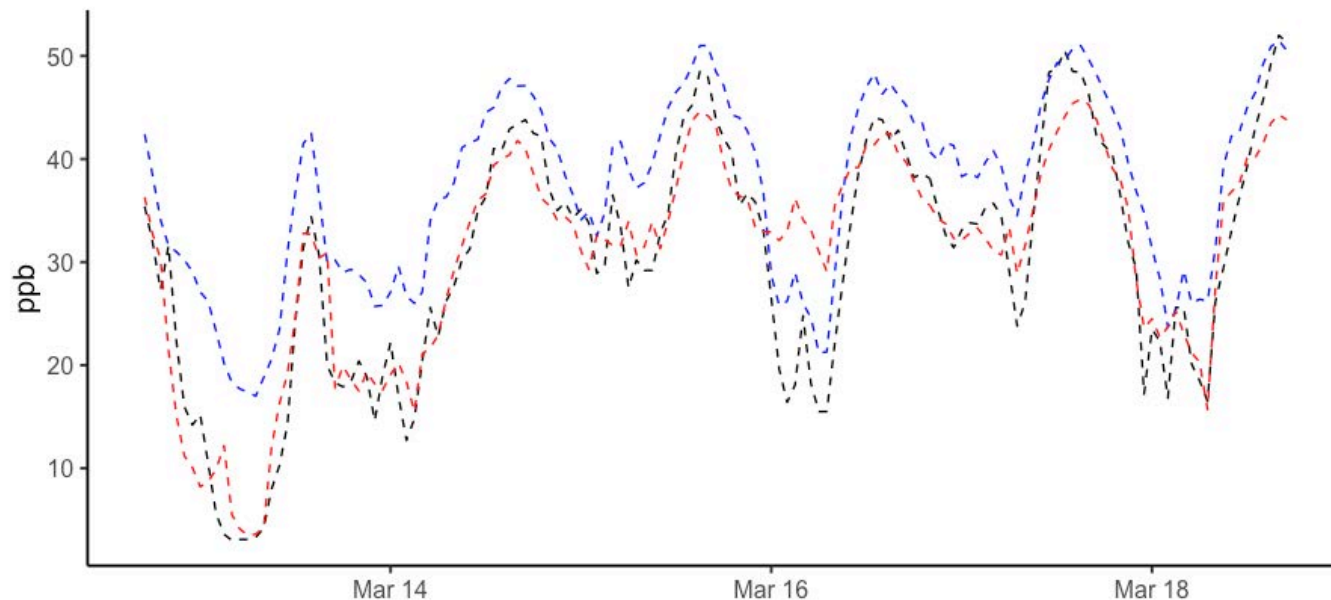
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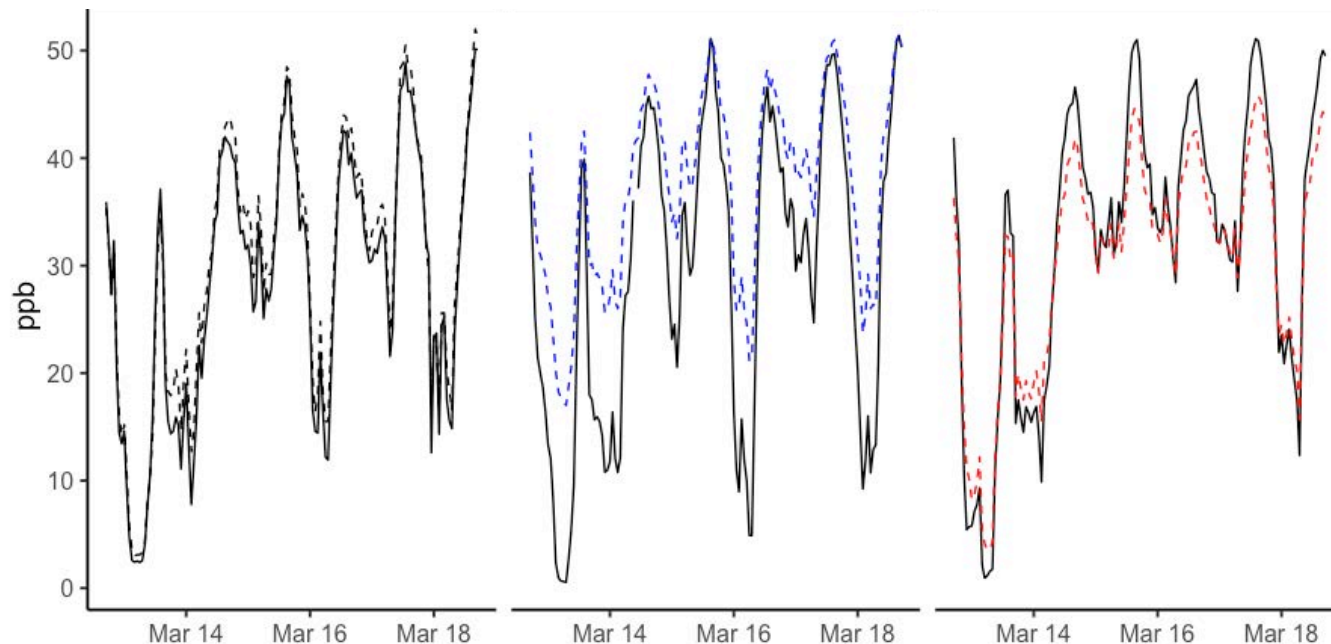
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Calibrating Sensor Devices



- Sensors show **reduced sensitivity** over time when not regularly calibrated.
- Calibration often done using **field co-location**.
- Can be difficult to characterise changes in the data.

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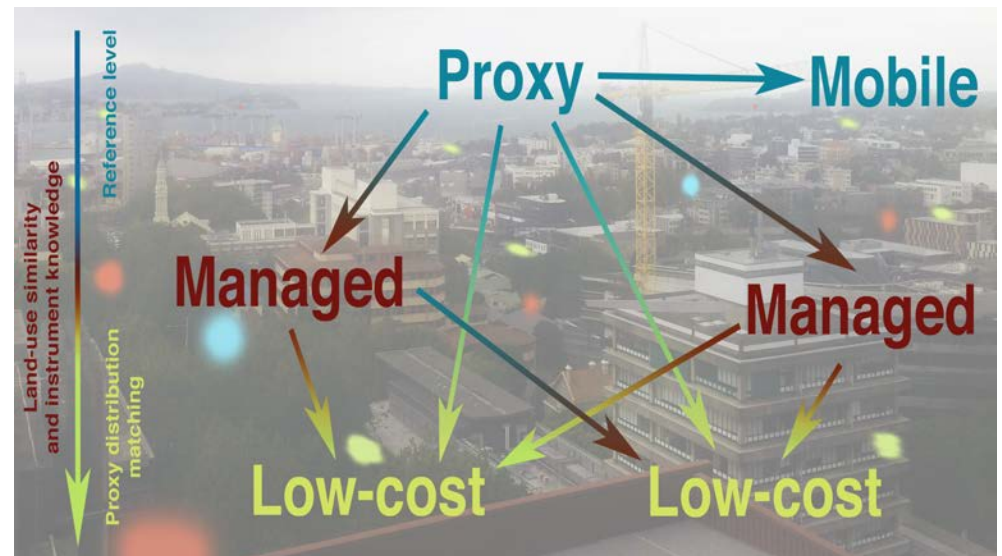
Remote Calibration

- Field visits can be time consuming and are unable to update continuously.
- Remote calibration – where **coefficients are found without co-location** – would resolve both these points.
- Need a method that is proven to find reliable coefficients.



Calibration Method

- Data as running three-day samples.
- High quality measurements provide ground-truth information – “proxies”.
- Use **land use similarity** to improve the proxy sample.
- Verify calibration by co-located regulatory measurements.



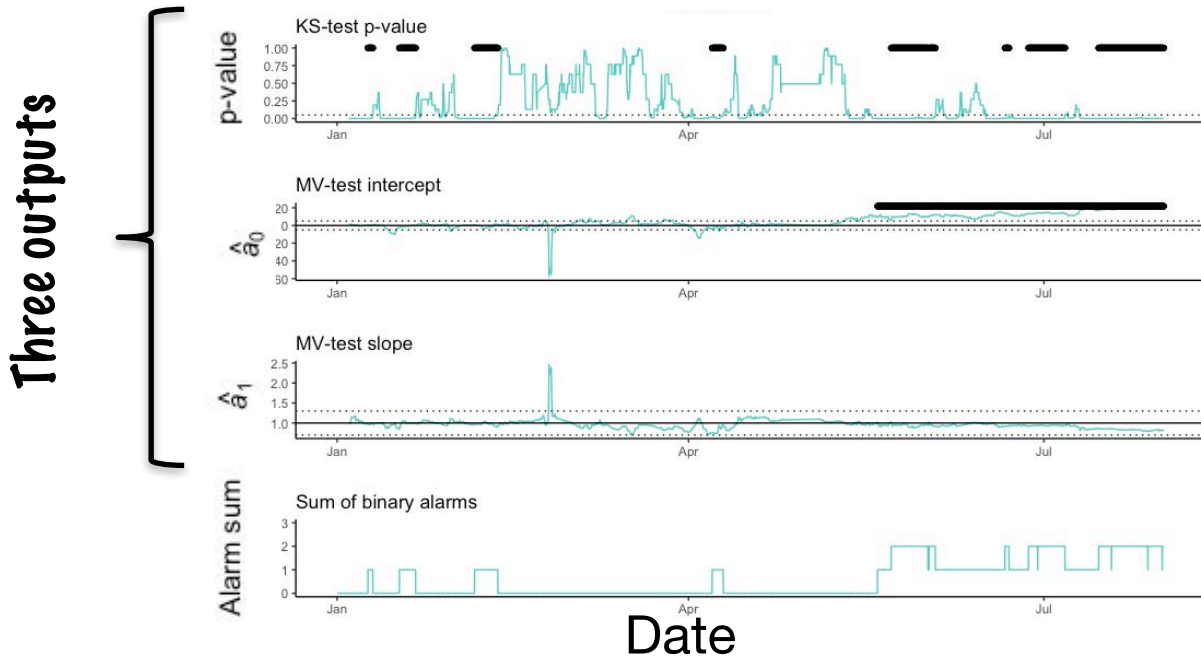
Calibration Method

- Assumption One: Sensor data (Y) is **linearly related to 'true' concentration** (X) over a sample of data.
- Assumption Two: Selected remote proxy data (Z) \sim X.
- Reworking assumptions gives the remote calibration coefficients:

$$\hat{a}_1 = \sqrt{\frac{\sigma^2 \langle Z(t) \rangle}{\sigma^2 \langle Y(t) \rangle}}$$

$$\hat{a}_0 = \mu \langle Z(t) \rangle - \hat{a}_1 \mu \langle Y(t) \rangle$$

Calibration Method



- Two tests with three outputs to check sensor data.
- Thresholds defined where alarm signalled.
- Consistent signalling less likely from natural variability.
- If >1 outputs with consistent alarms, the calibration used.



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Article

Solution to the Problem of Calibration of Low-Cost Air Quality Measurement Sensors in Networks

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Article

Data Verification Tools for Minimizing Management Costs of Dense Air-Quality Monitoring Networks

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Miskell et al. *ACS Sens.* 2018, 3, 832-843

Miskell et al. *Env. Sci. & Technol.* 2016, 50(2), 835-846

Data

Data from two networks measuring ozone (ppb) by gas-sensitive semiconducting sensors manufactured by Aeroqual.

Network 1: Vancouver, Canada

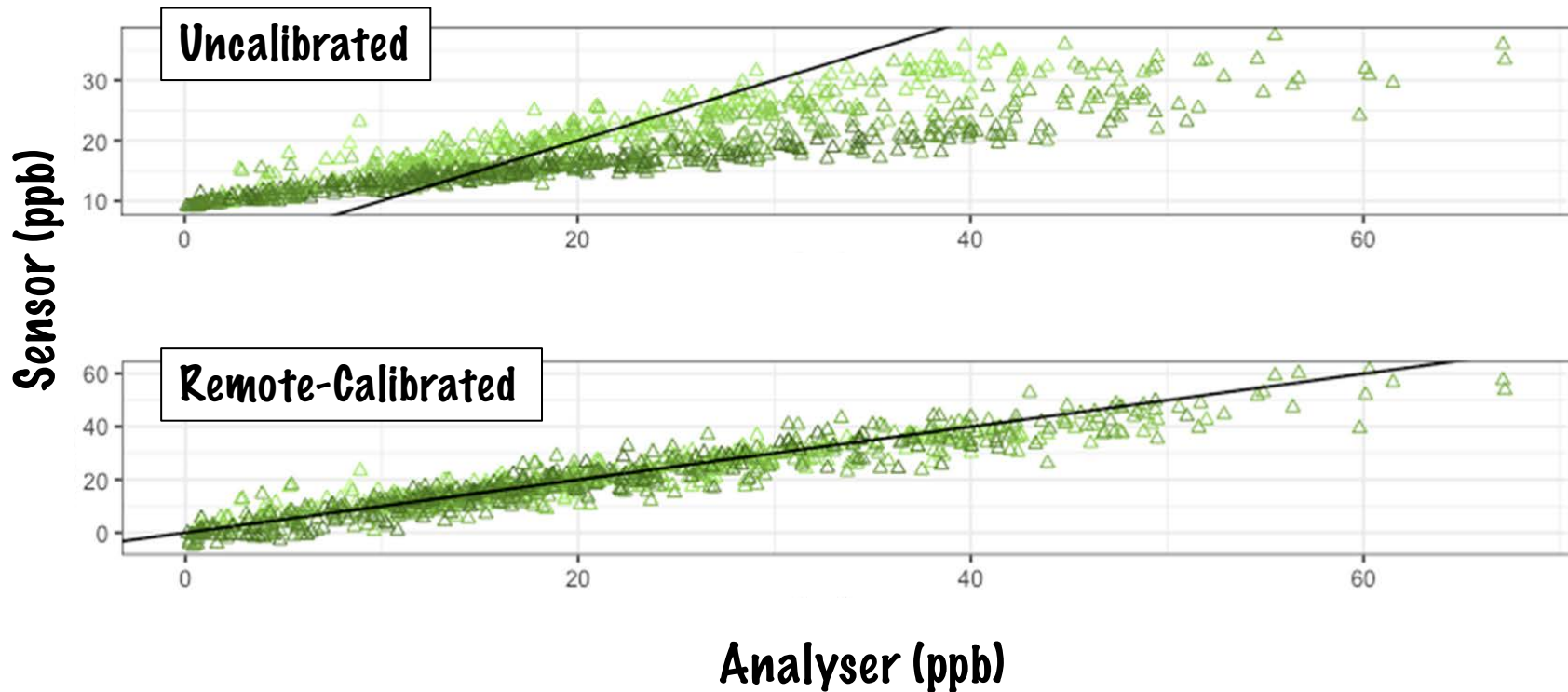


Network 2: Los Angeles, California



Vancouver

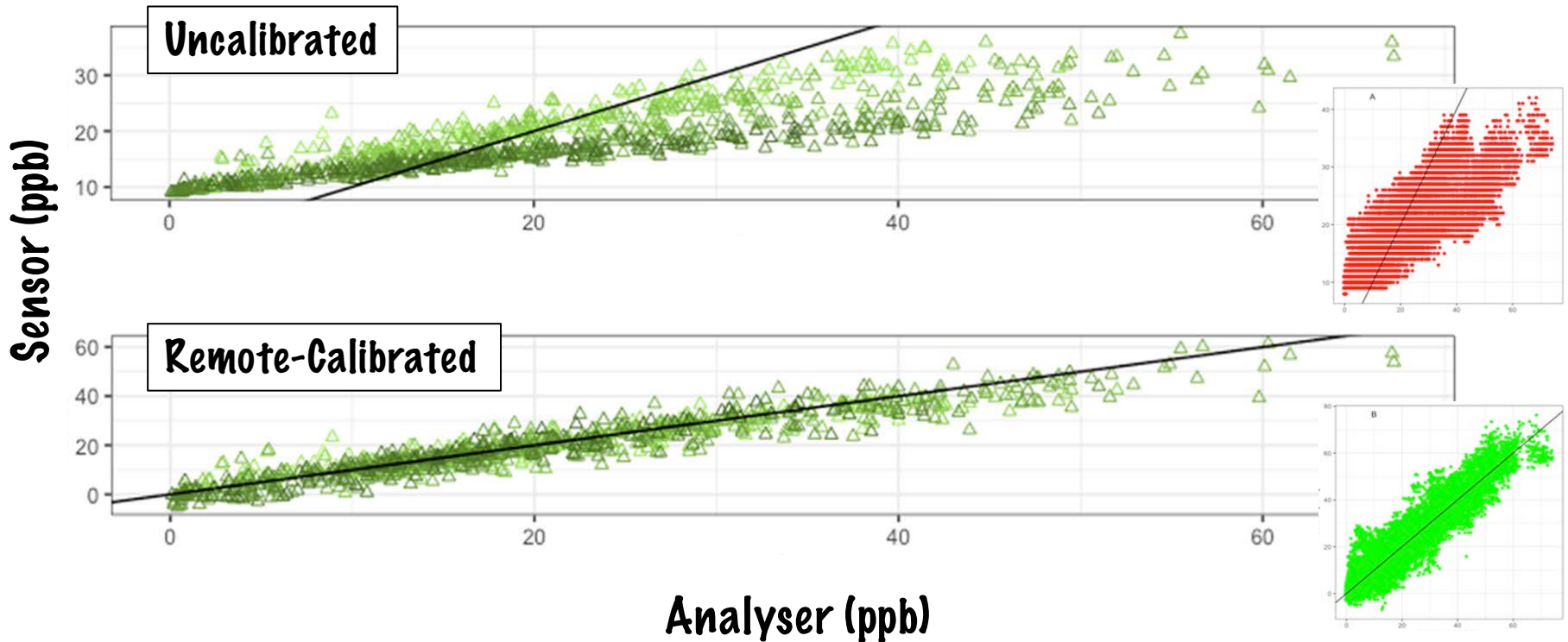
1-hour data.



Vancouver

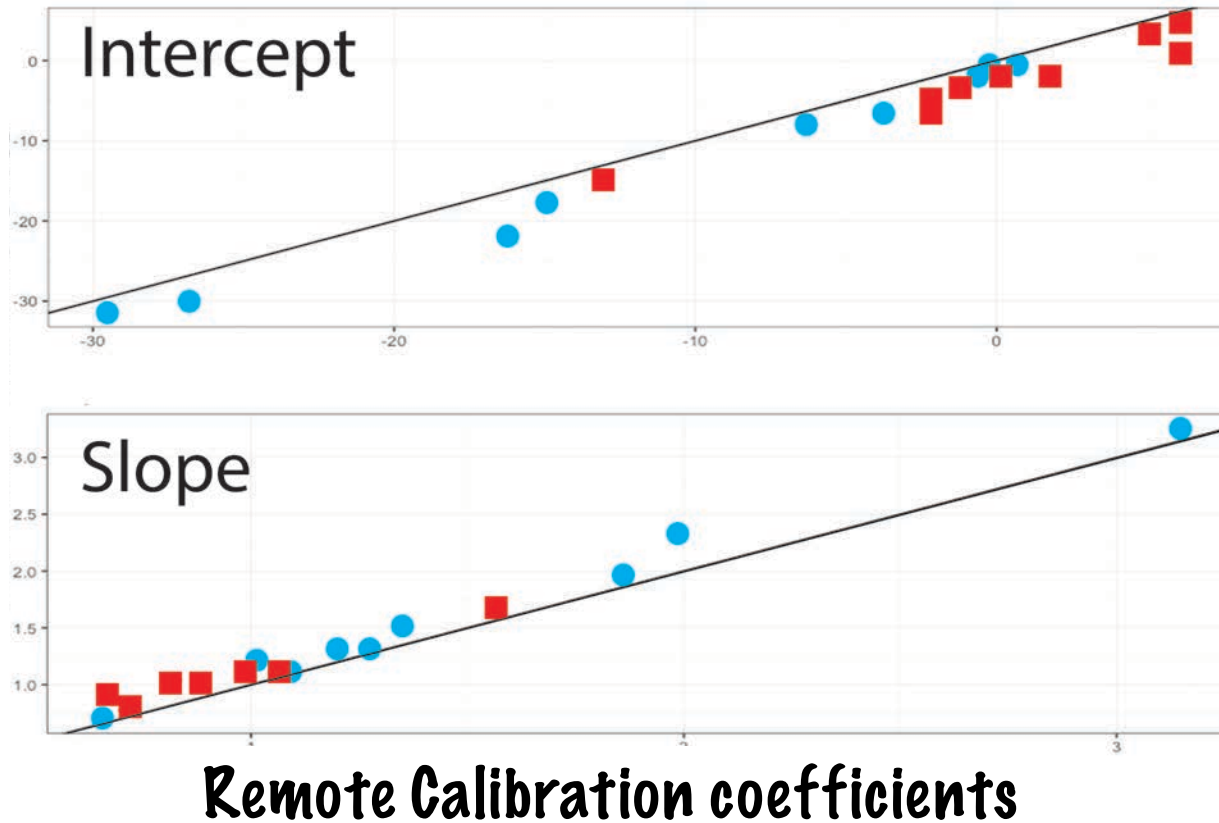
1-hour data.

1-min
data.



Vancouver

Field co-location coefficients

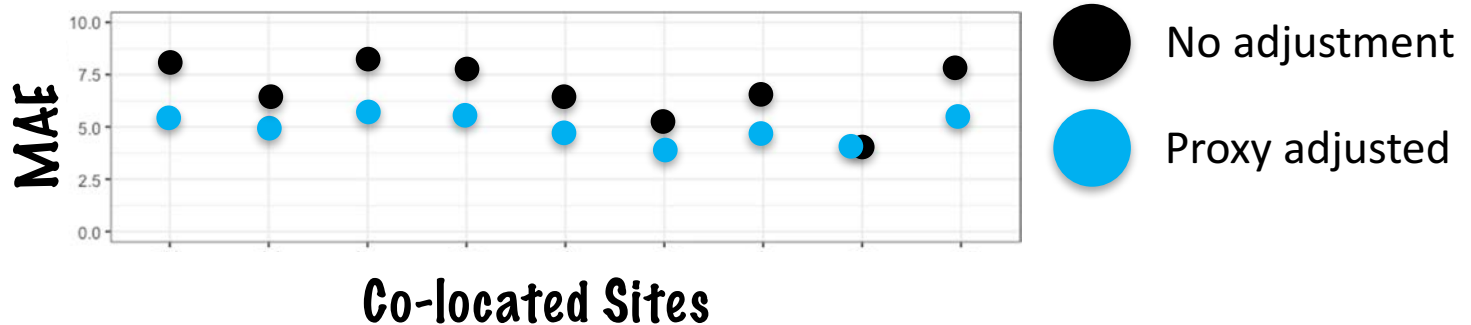


 First week of monitoring

 Last week of monitoring

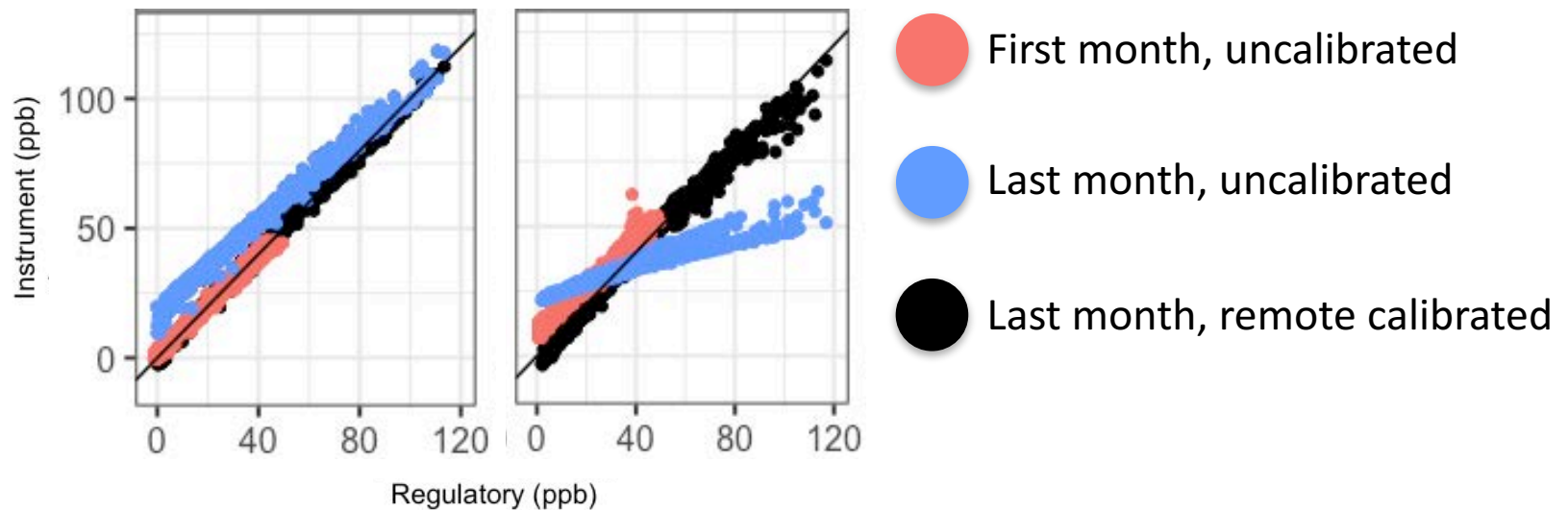
Vancouver

Sensor land use	Proxy land use		
	Urban	Suburban	Rural
Urban	1	0.76	0.68
Suburban	1.23	1	0.87
Rural	1.28	1.03	1



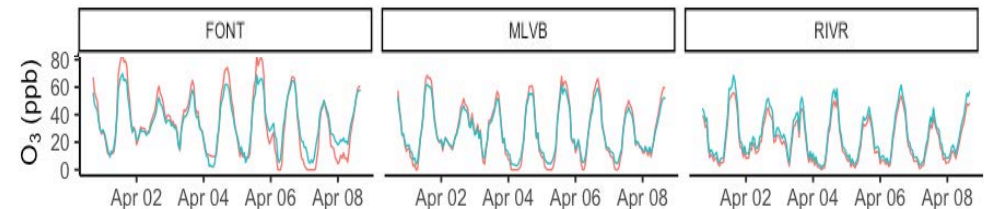
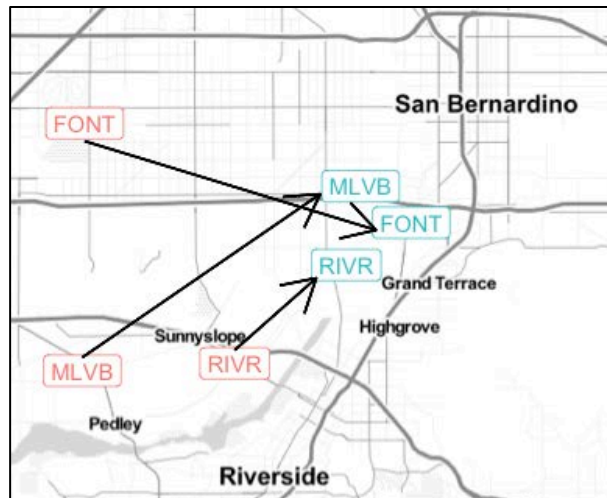
Land use adjustments improved the proxy when it was a poor fit.

Los Angeles



- Network of ~ 100 sensor devices.
- Remote calibration improved accuracy of data (O_3).

Los Angeles



- Field-calibrated sensors moved to non-co-located sensor devices calibrated by the remote method.
- Time-series showed good agreement between the two devices.

Summary

Remote calibration:

- Cost-effective.
- Check and update any time.
- Do not require regular access to sites.
- Shown to work for ozone sensor devices in two networks.

- Finding appropriate proxies a challenge.
- Unknown level of uncertainty added to the measurement.

Thanks for listening!

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SCIENCE



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HIKINA WHAKATUTUKI

CallaghanInnovation

aeroqual 

Thanks to **Andrea Polidori, Vasileios Papapostolou, Brandon Feenstra & Berj Der Boghossian** from South Coast Air Quality Management District, and to **Ken Reid & Julie Saxton** from Metro Vancouver for regulatory data and access to sites.

Thanks to the many individuals and facilities who generously hosted instruments!