

Identifying spatiotemporal patterns in air quality: applications in source apportionment

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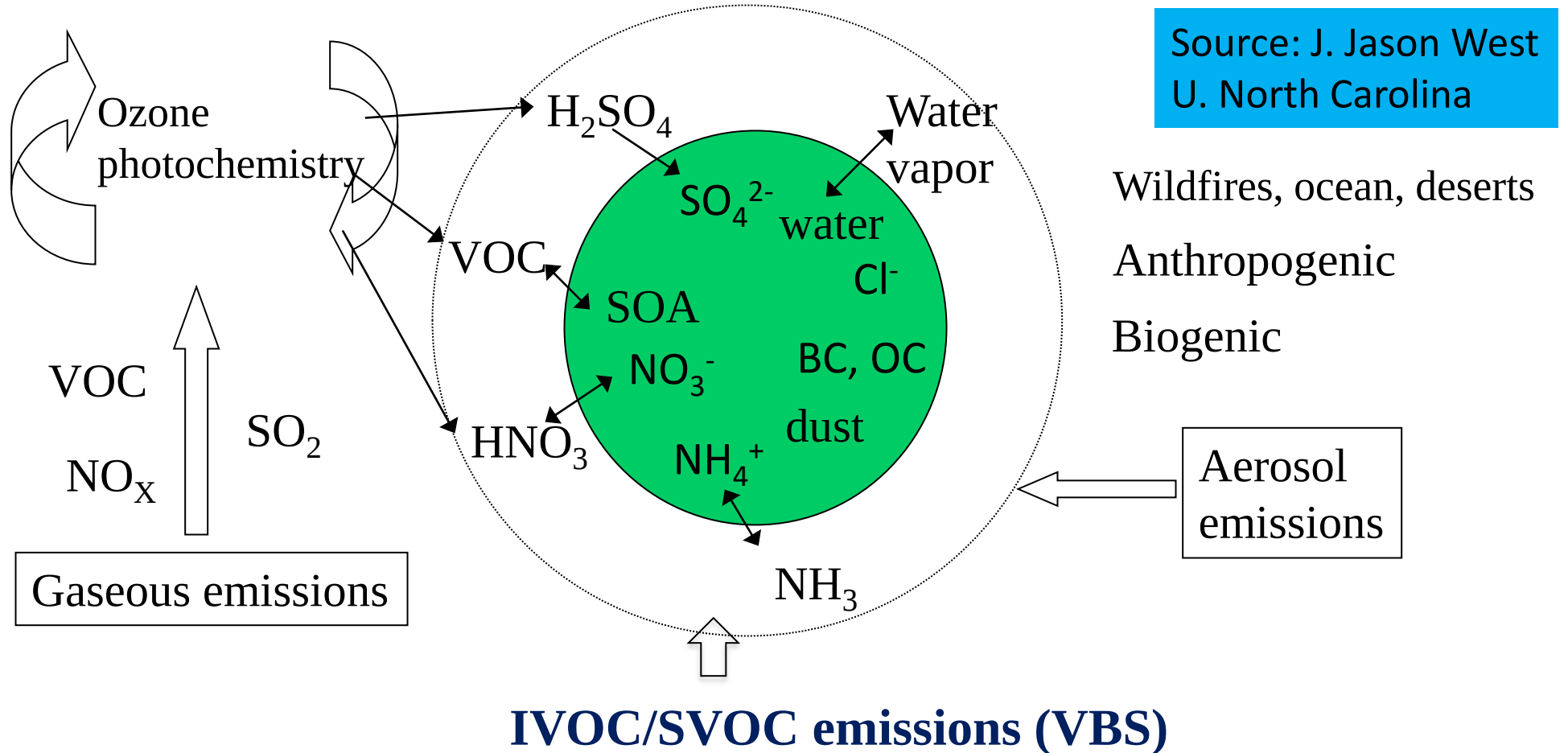


ACKNOWLEDGEMENTS

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Introduction: how to handle air pollution?

“Too many moving parts”



Which emission controls are better?

Introduction: Source Apportionment approaches

- Air Quality Modeling (AQM): applicable to all pollutants, gases and particles
 - Problem: error propagation is hard to quantify
 - Advantage: generate spatiotemporal (STP) distribution of pollutants
- Receptor Modeling (RM): applicable only to particulate matter
 - Problem: expensive to carry out, cannot be applied retrospectively, only a few sites are characterized per city
 - Advantage: uncertainties are easier to estimate, results must obey the mass balance for tracers

A new approach: fuzzy clustering

- Assumption:
 - Total pollutant concentration = sum {source contributions}
- Traditional cluster analysis is a hard classification
- Outliers pose additional problems
- Fuzzy clustering approach:
 - Each observation 'i' may belong to one or more clusters 'j'
 - Probability is the membership function u_{ij}
 - FC is mass preserving by design (partitioning)
 - Outliers are explicitly handled

Fuzzy Spatiotemporal apportionment (FUSTA)

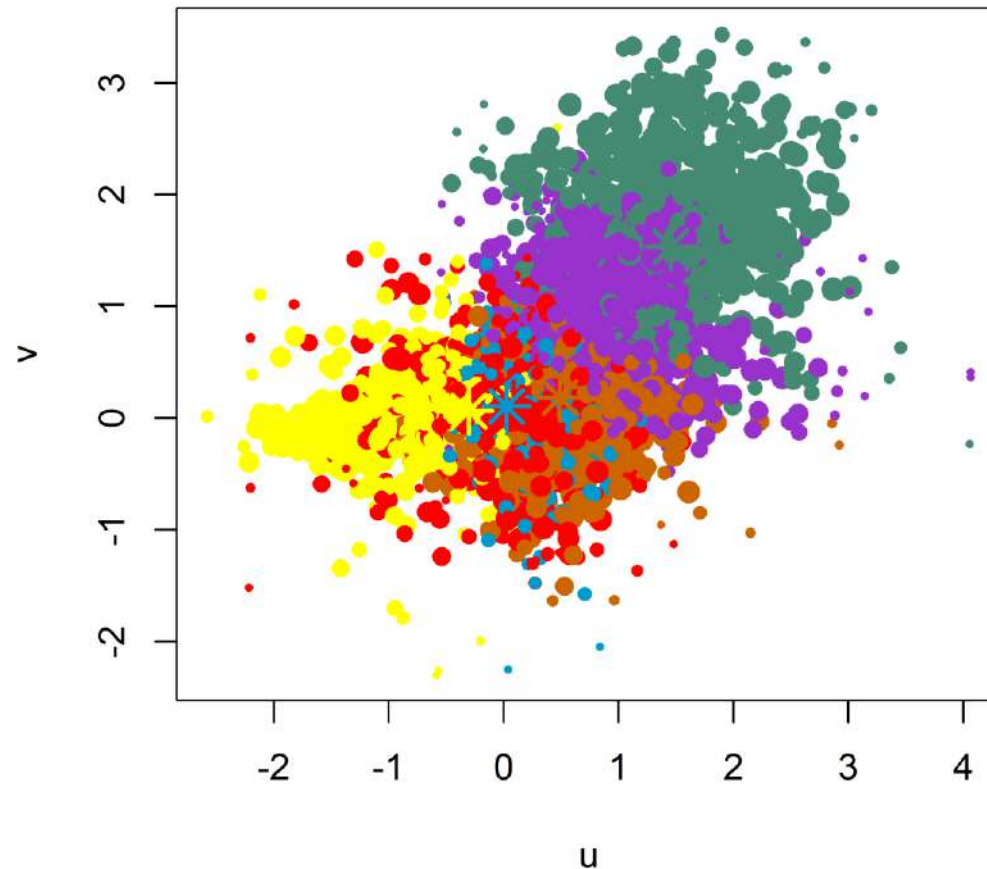
- Observations: $X=\{\text{SO}_2/\text{PM}_{10}/\text{PM}_{2.5}, \text{w. speed, w. direction, temp., press.}\}$

$$\min_{U,C} J_{FKMNE} = \sum_{i=1}^n \sum_{k=1}^p u_{ik} \cdot \|x_i - c_k\|^2 + t \cdot \sum_{i=1}^n \sum_{k=1}^p u_{ik} \cdot \log(u_{ik}) + \sum_{i=1}^n \delta^2 \left(1 - \sum_{k=1}^p u_{ik} \right)^2$$

- *s.t.* $u_{ik} \in [0,1]; \sum_{k=1}^{p+1} u_{ik} = 1$
- Computations are carried out using package *fclust* in R environment (Ferraro and Giordano, 2019)

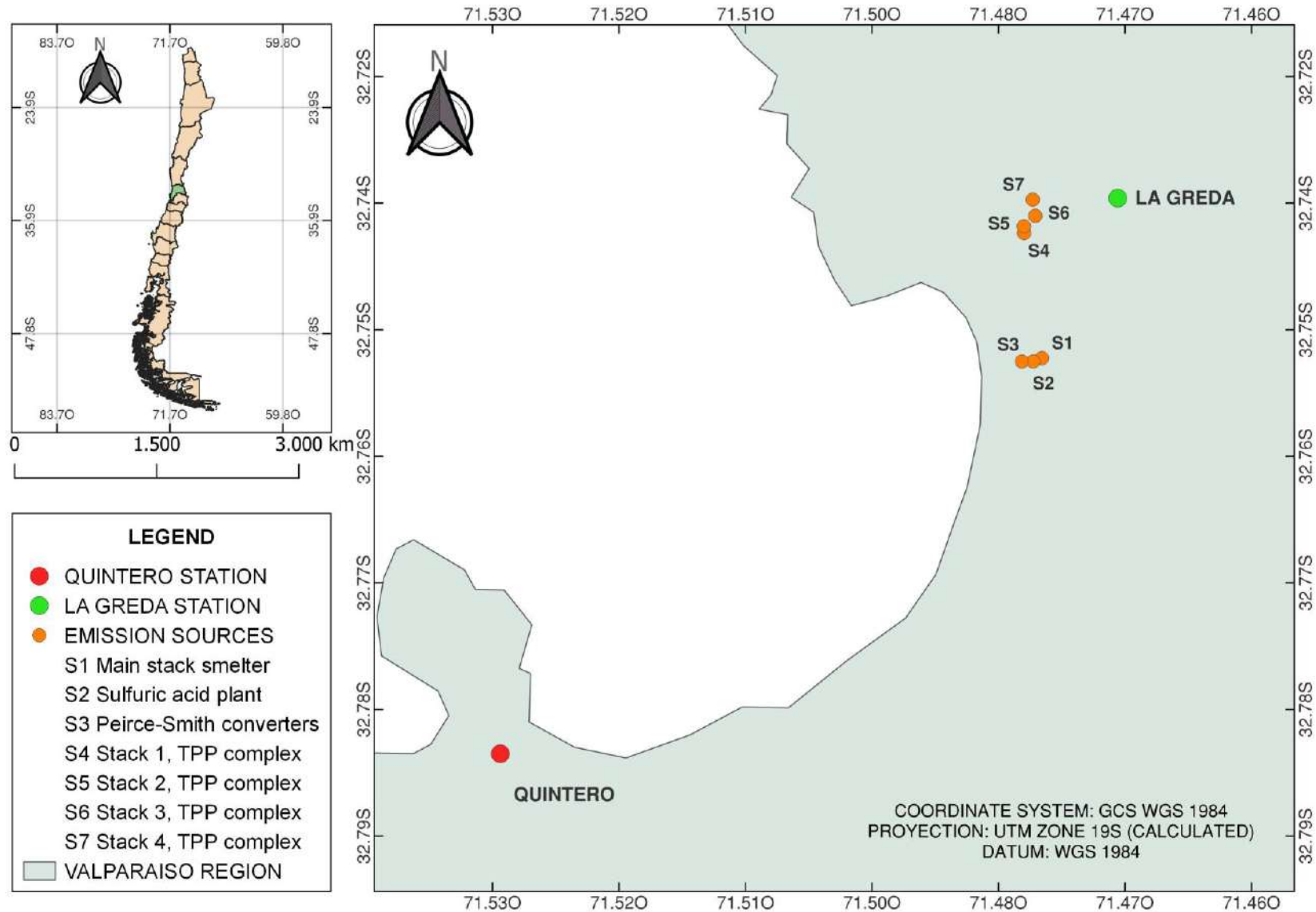
Example of application with $p=5$

Projection on (u,v) space



Symbol magnitudes
proportional to
the membership
values.
Noise cluster is
implicit.

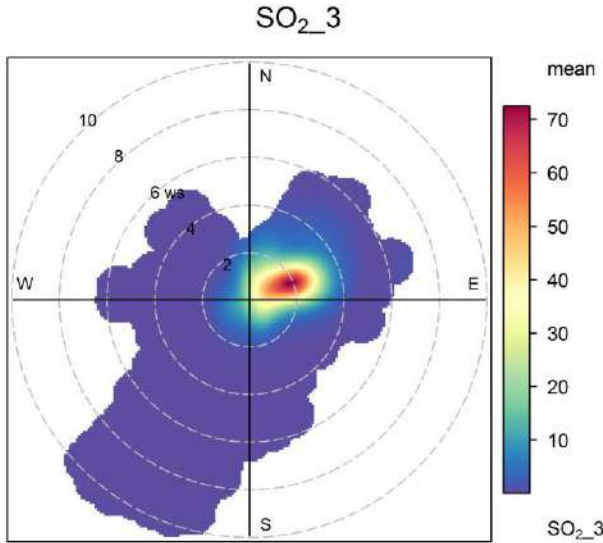
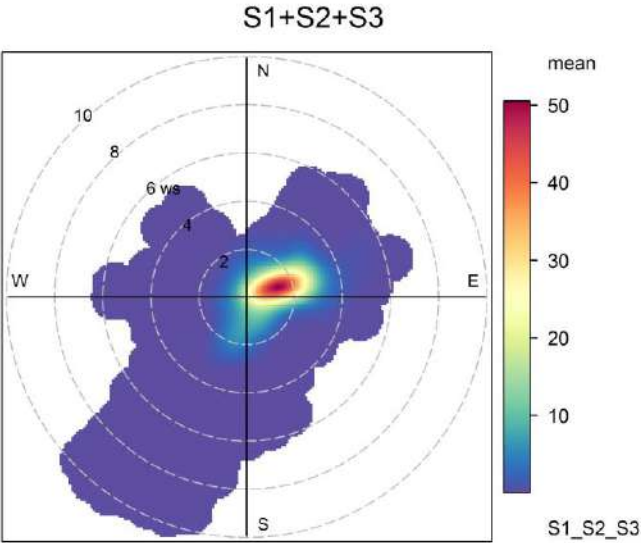
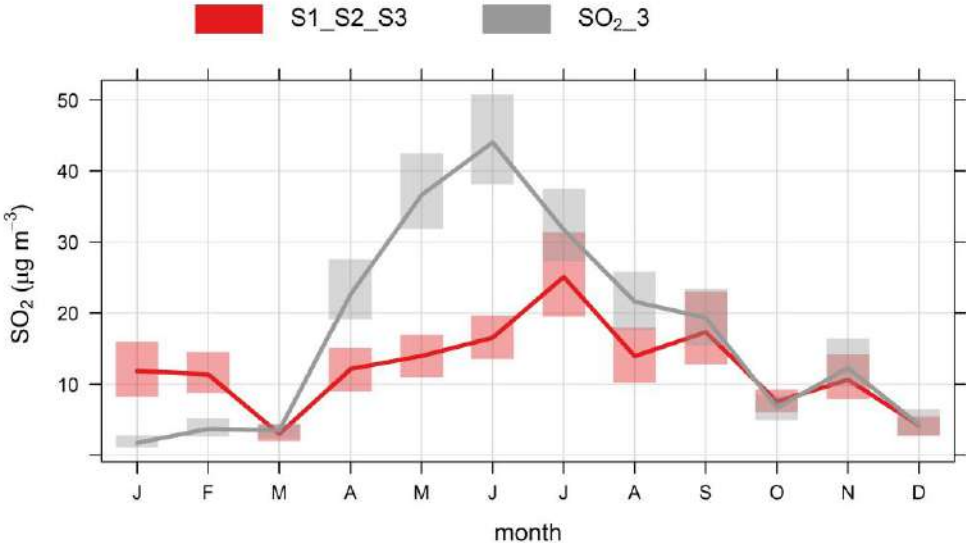
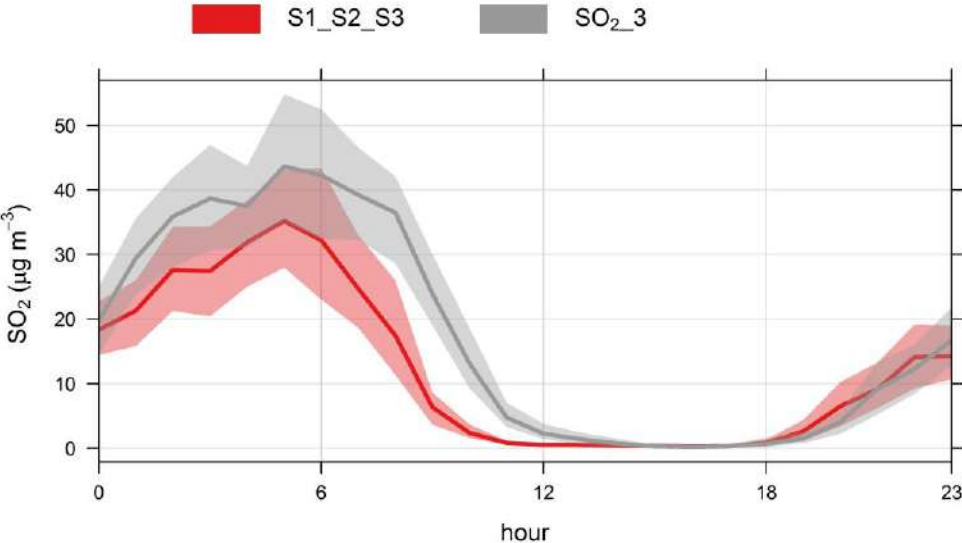
Case study 1: industrial area



Jorquera H., Villalobos A.M.
Journal of Hazardous Materials,
443 B, 130335, 2023.

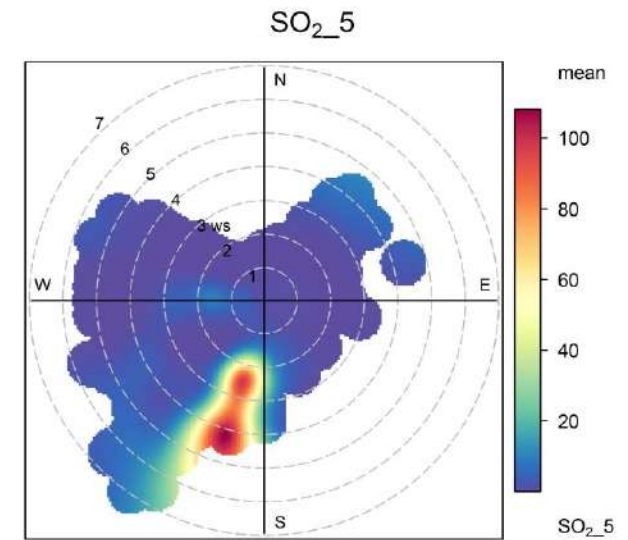
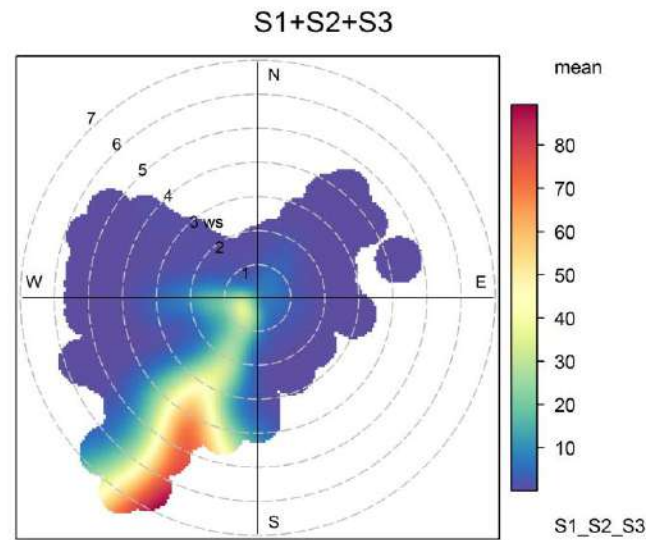
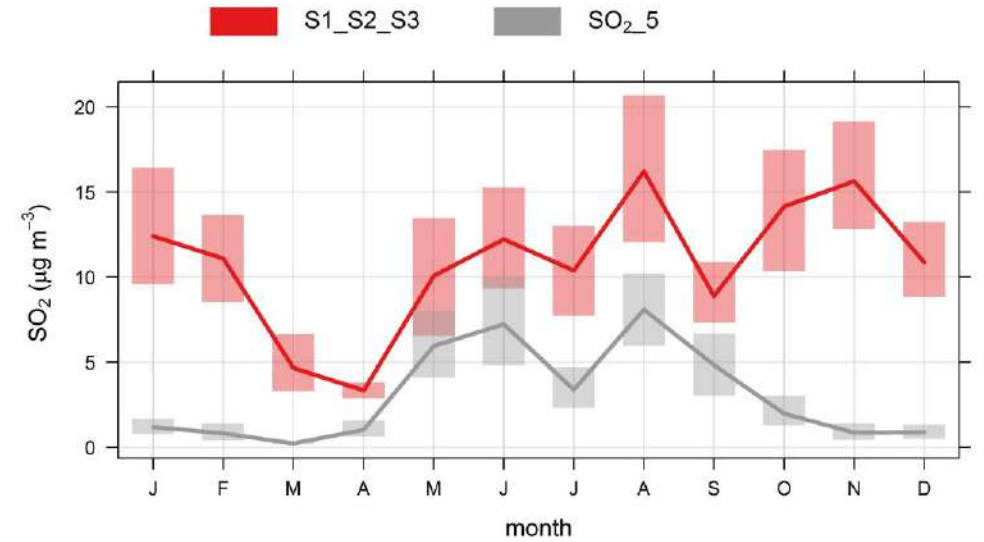
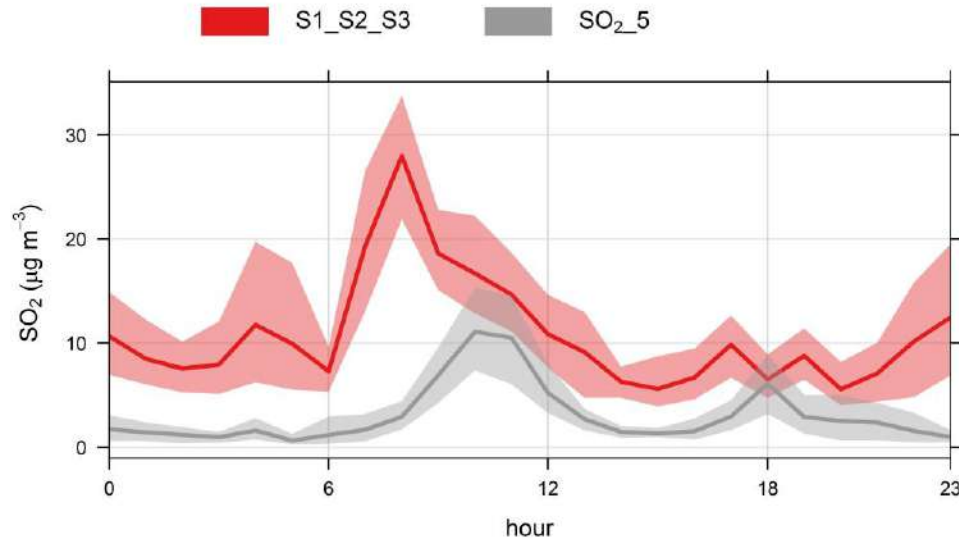
Case study 1: predicted contributions of copper smelter on red receptor

AQM: WRF-CALPUFF+
CEMS emissions

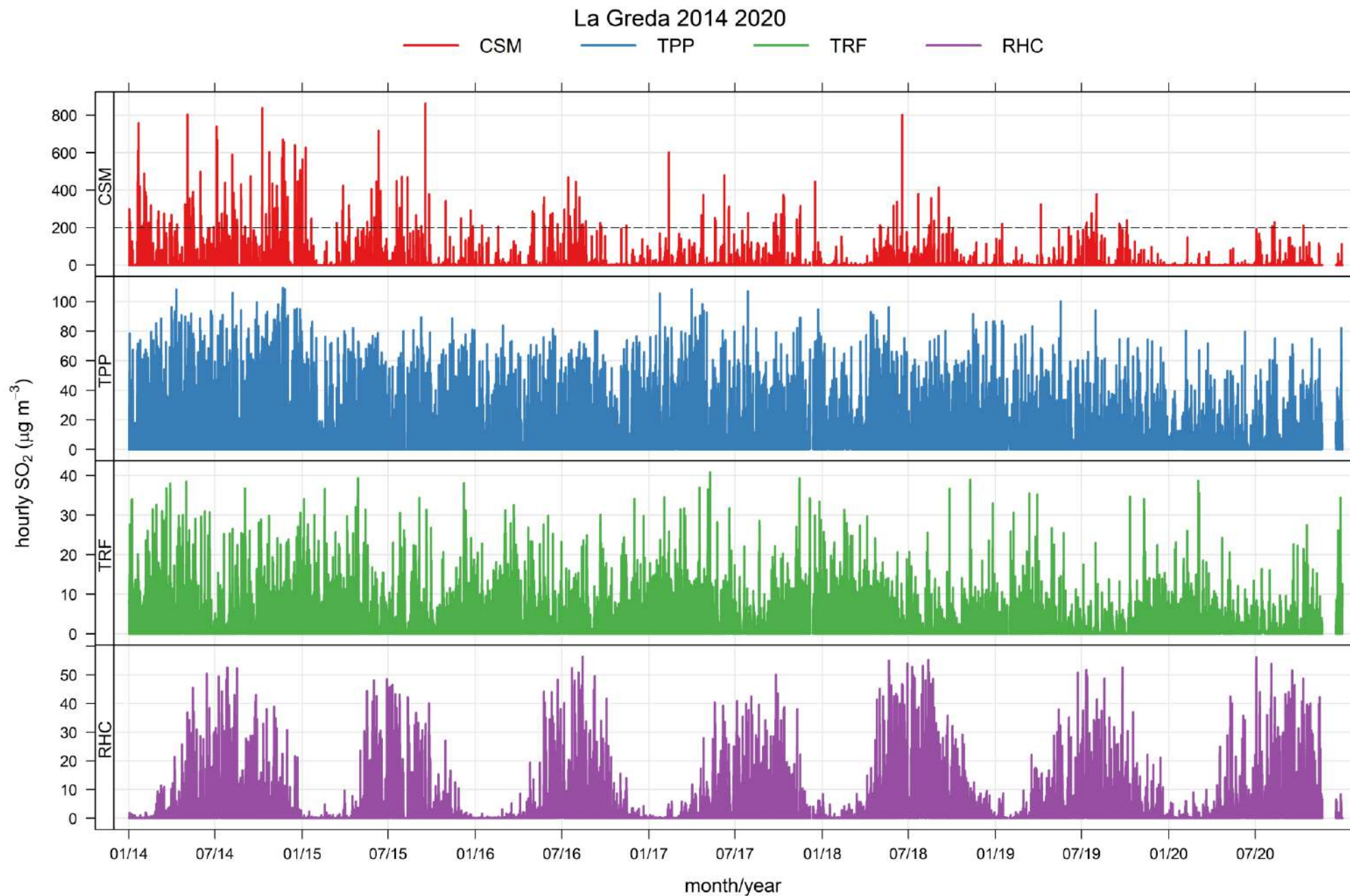


Case study 1: predicted contributions of copper smelter on green receptor

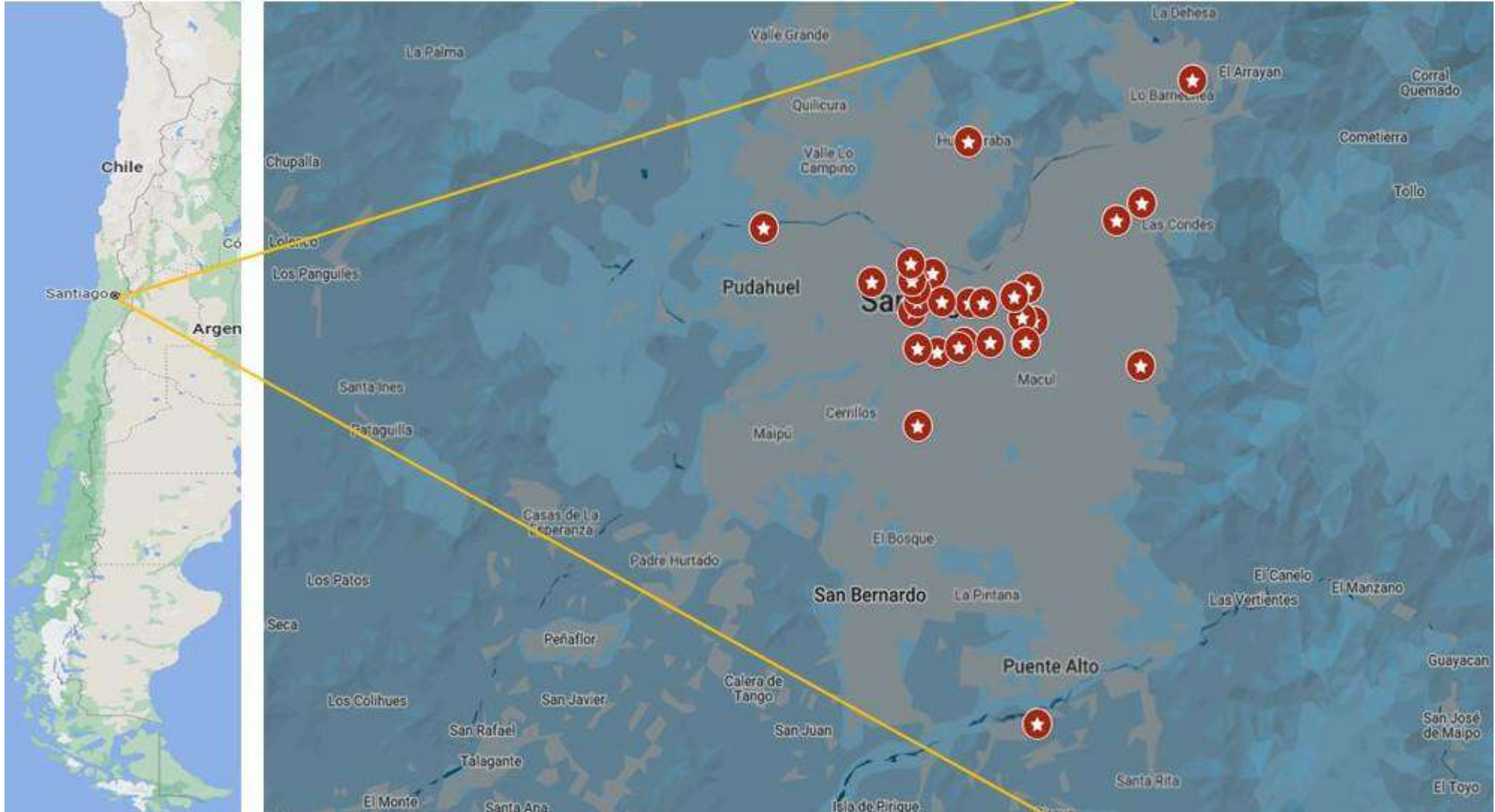
AQM: WRF-CALPUFF+
CEMS emissions



Long-term SO₂ SA



Case study 2: outdoor/indoor PM_{2.5} in schools



S. Nourani and H. Jorquera, Environmental Science - Processes & Impacts, 2024, december

Monitoring campaign

30 naturally ventilated schools

The measurement was carried out with low-cost sensors.

Temperature, relative humidity (RH) and CO₂ concentration were measured using HOBO® CO2 loggers.



S5- Far classroom



S11- Outdoor



PurpleAir Inc, UT, USA

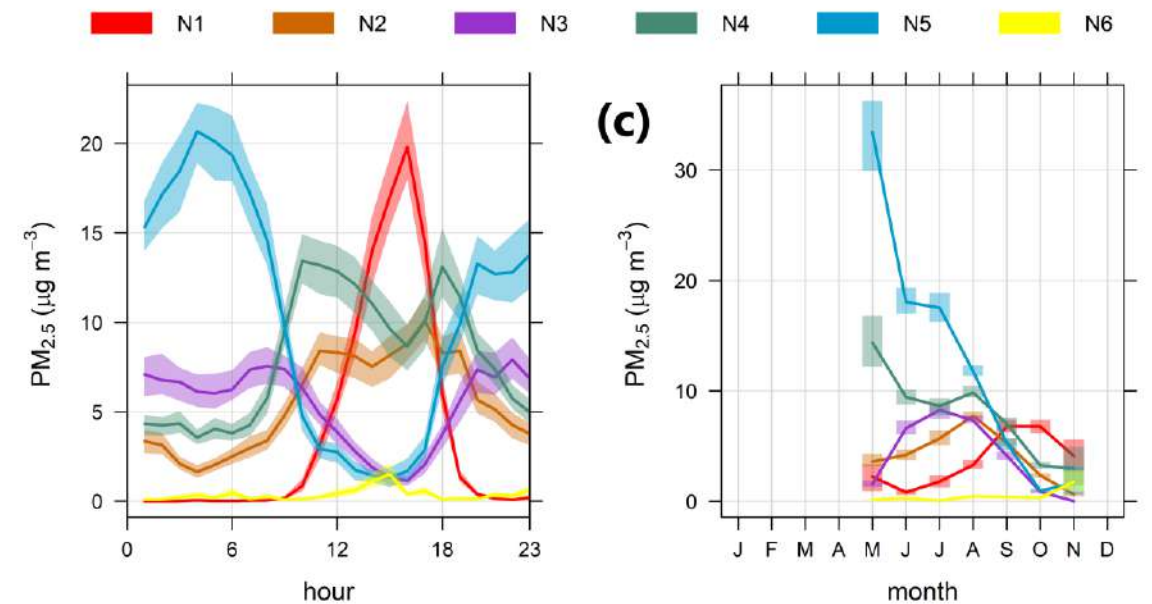
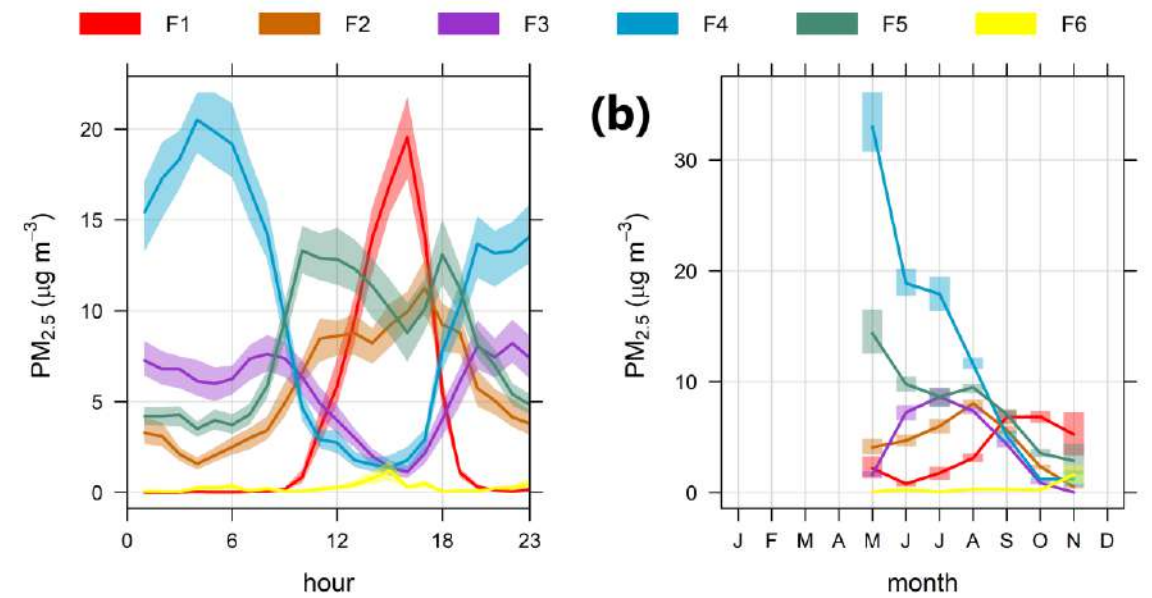
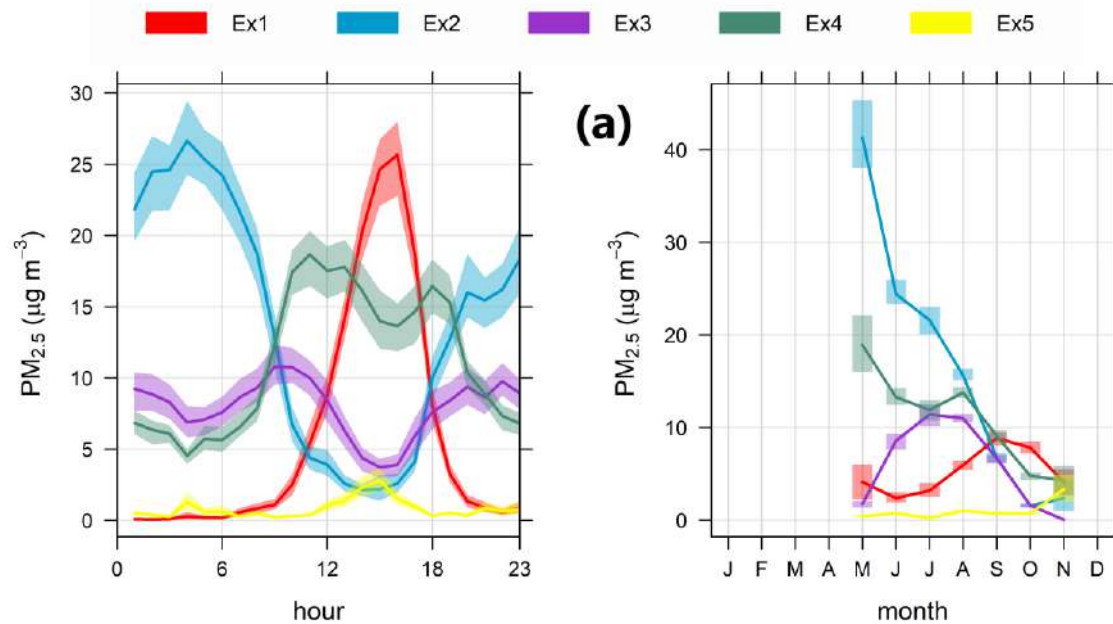


Onset HOBO MX1102A

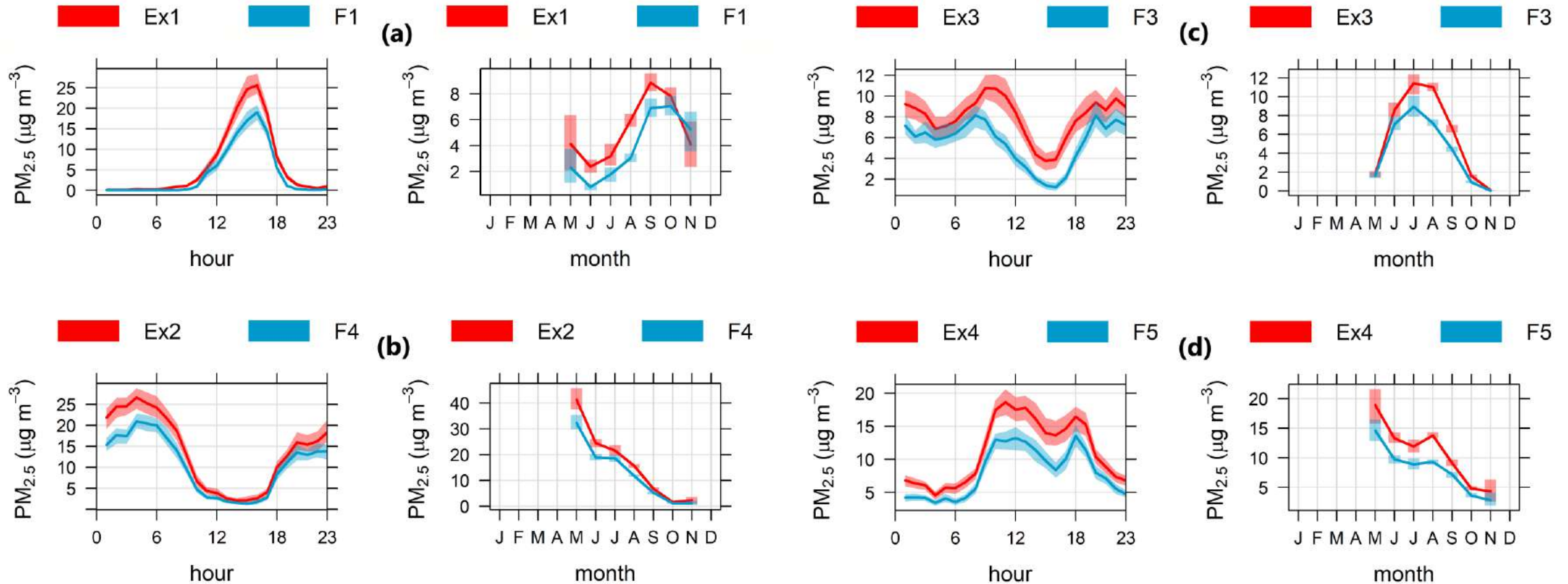


S30- Near classroom

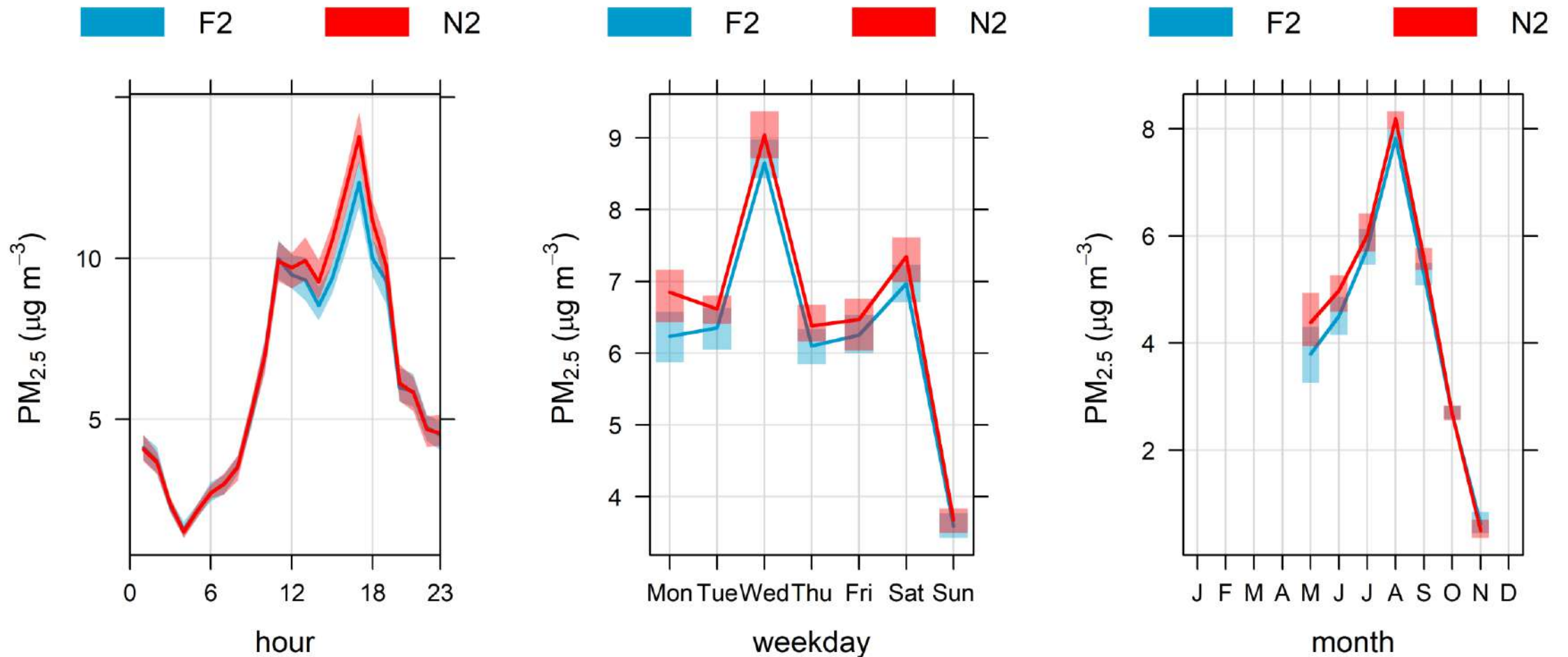
Case study 2: STP found in outdoor (a) and classrooms (b), (c)



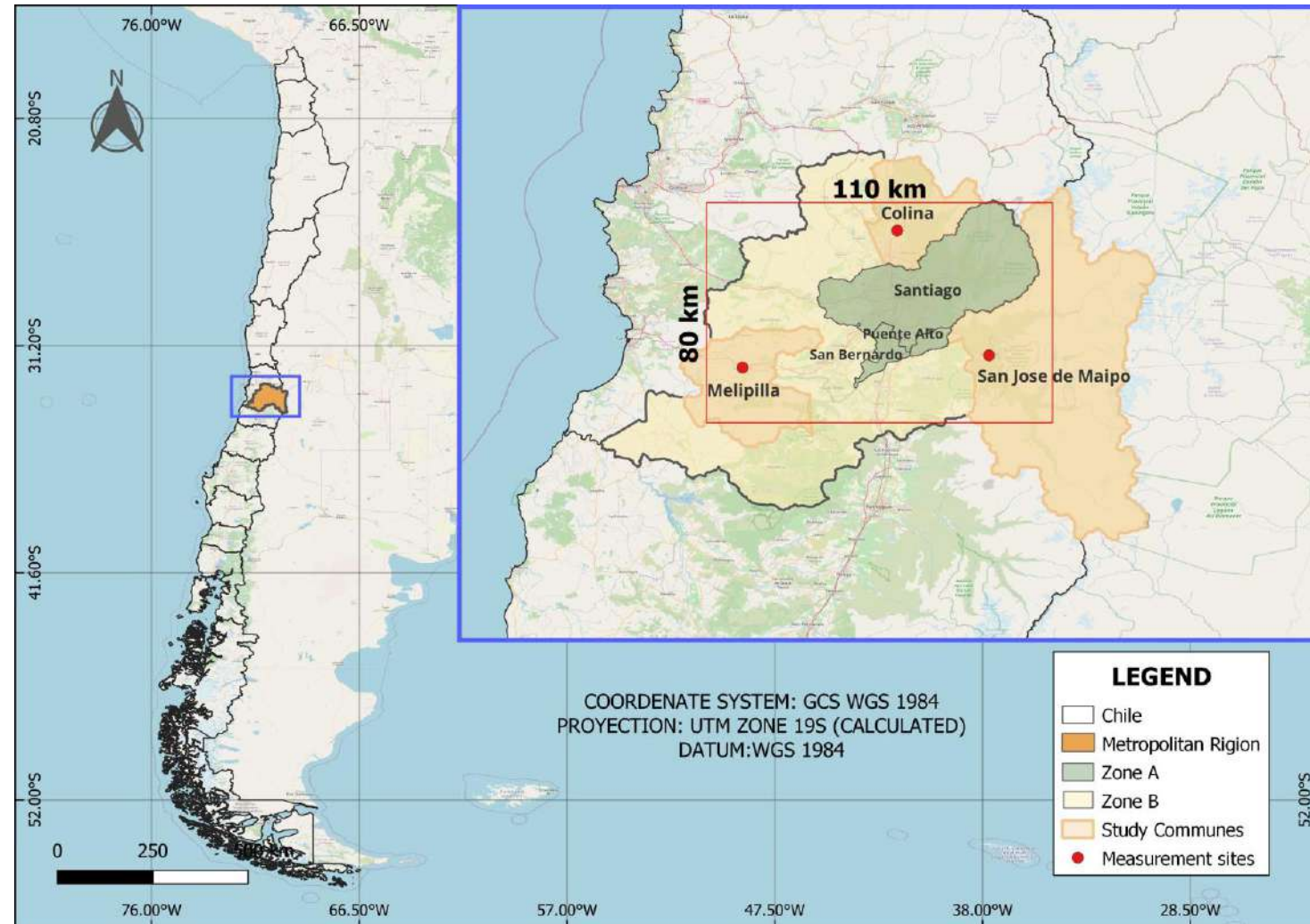
Case study 2: outdoor/indoor PM_{2.5} by cluster



Case study 2: indoor generated PM_{2.5}

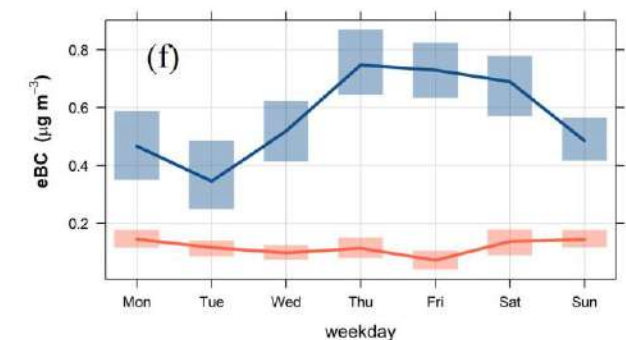
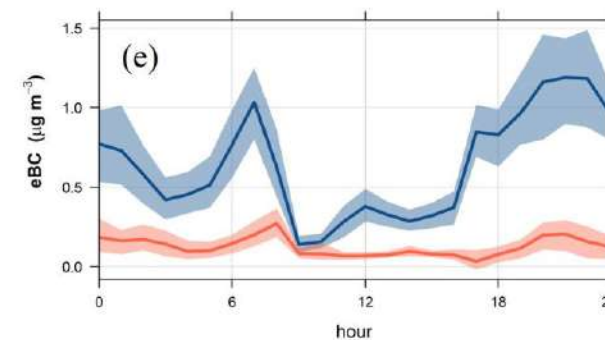
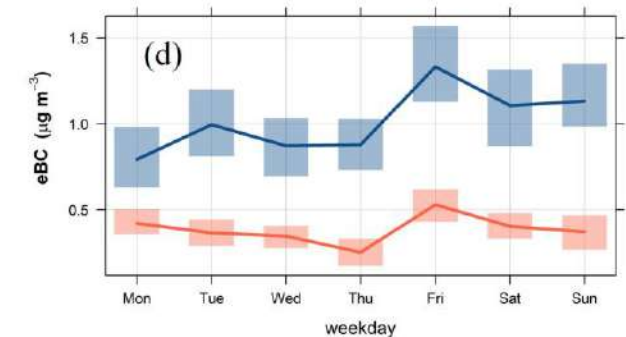
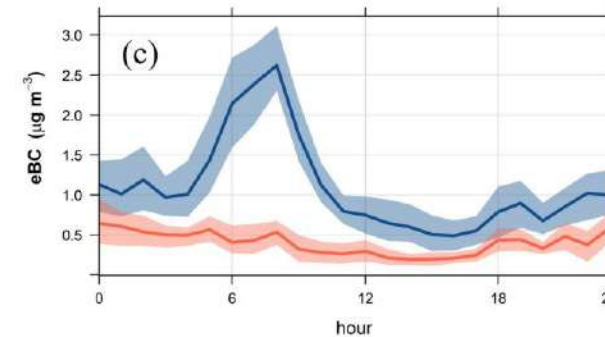
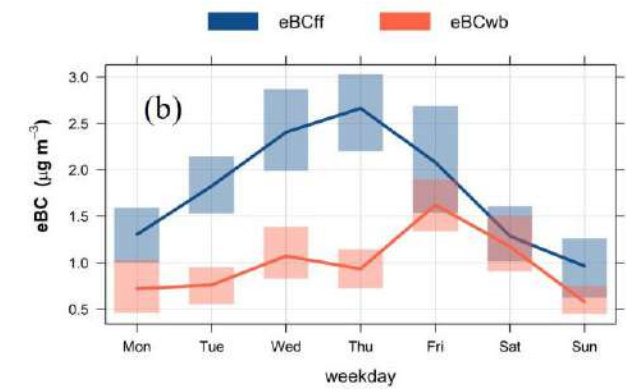
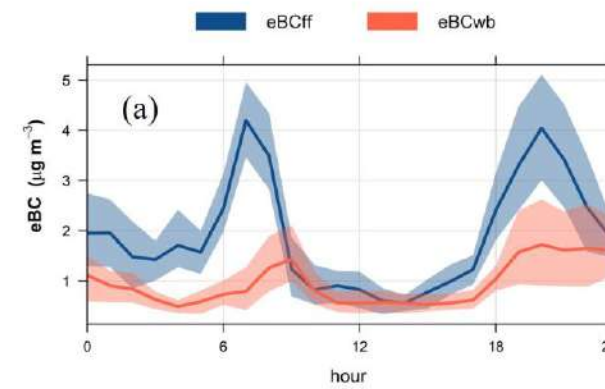


Case study 3: estimating local and non-local BC contributions in three suburbs of Santiago, Chile

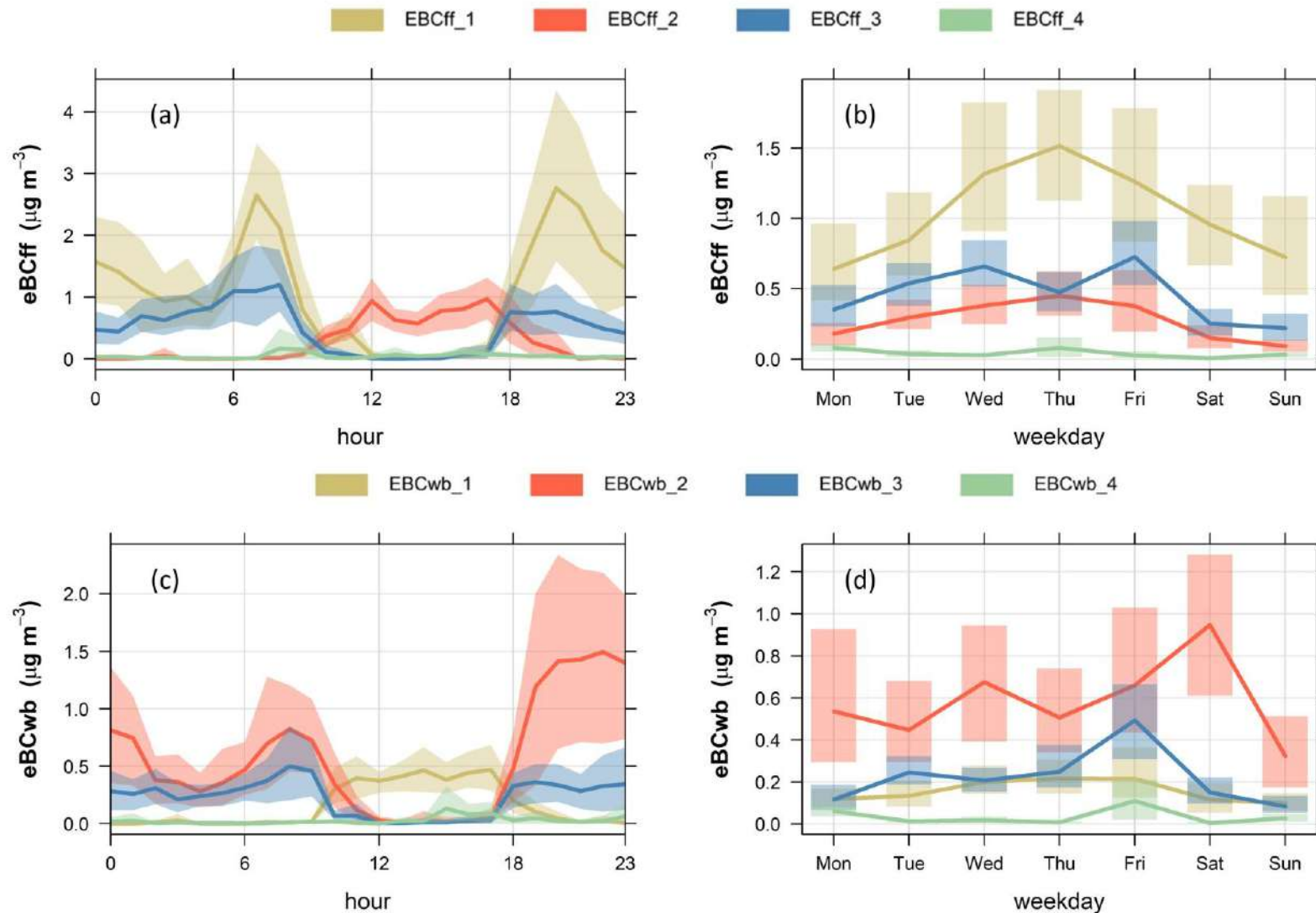


J. Rodríguez, AM Villalobos,
J. Castro-Molinare, H. Jorquera,
Environmental Pollution,
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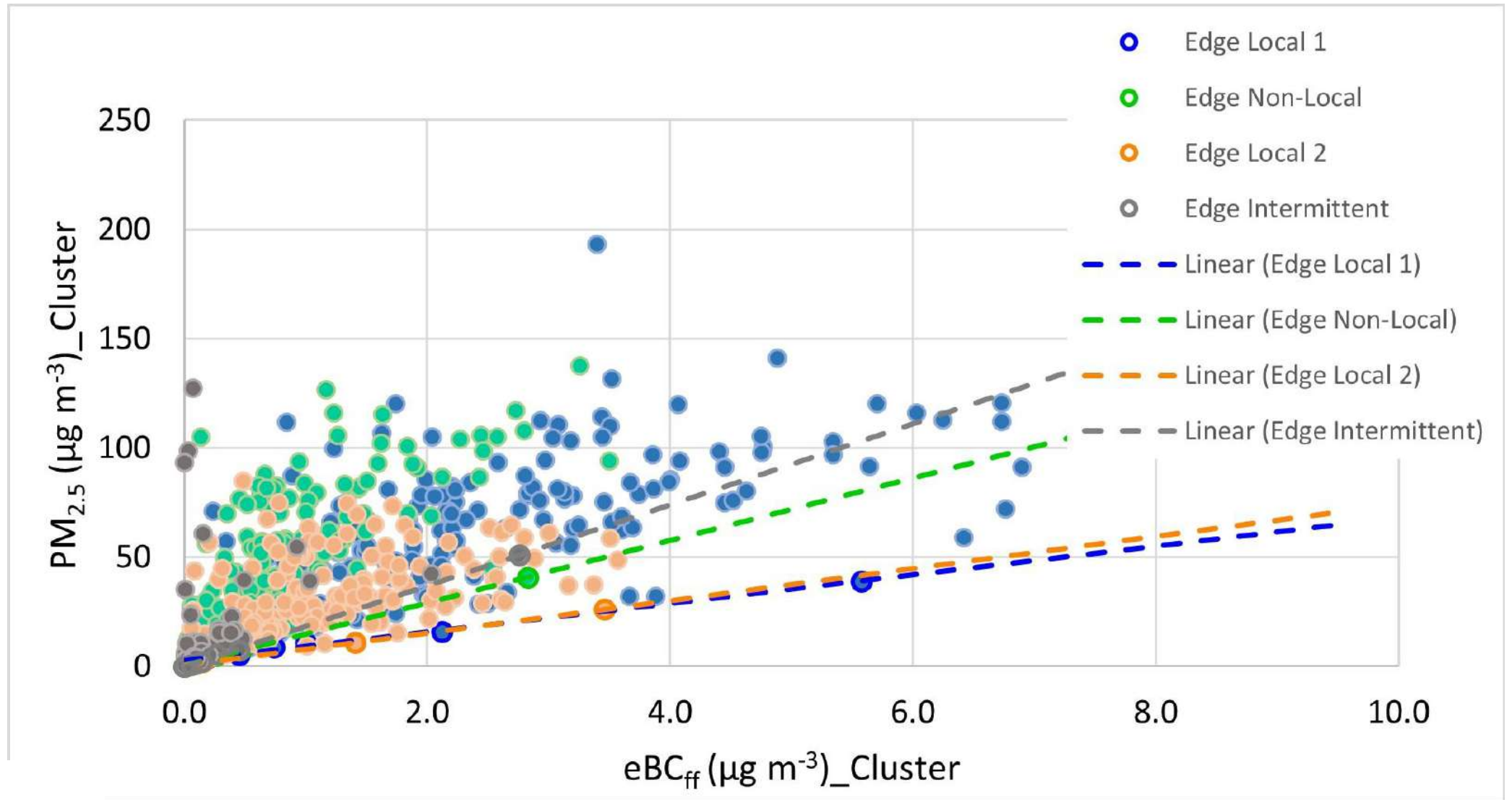
Case study 3: standard BC source contributions by applying aethalometer model to multiwavelength optical absorption data



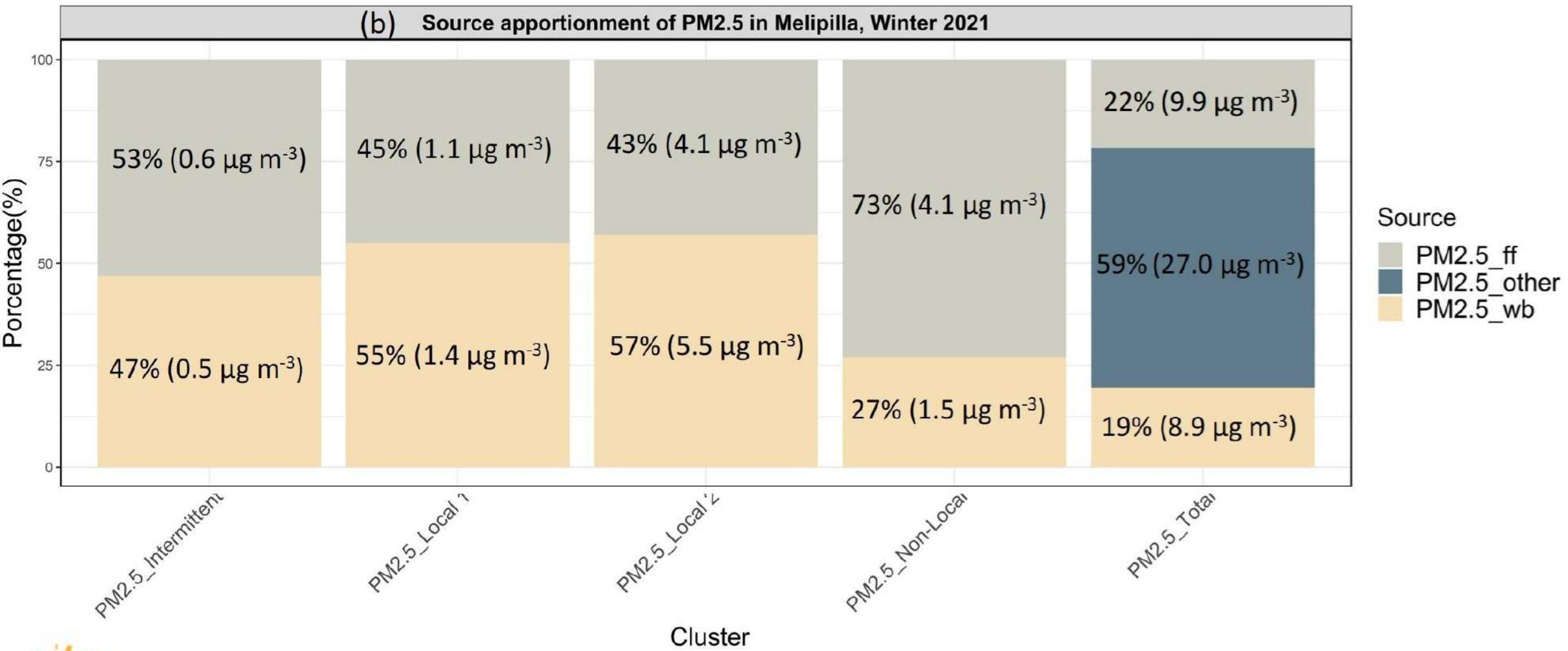
Case study 3: FC applied to eBC fractions (Colina site)



Case study 3: Implications for tracer method



Case study 3: Source identification results



CONCLUSIONS

- FUSTA is a new data-driven method for source identification
- It only requires ambient data for a pollutant and meteorology
- The method can handle pollutant outliers
- Applications include:
 - Source apportionment of gaseous pollutants
 - Discriminating among local and non-local pollution sources
 - Estimating indoor-generated PM contributions
- Results may be used to constrain specific emission sources
 - By using AQM without boundary conditions