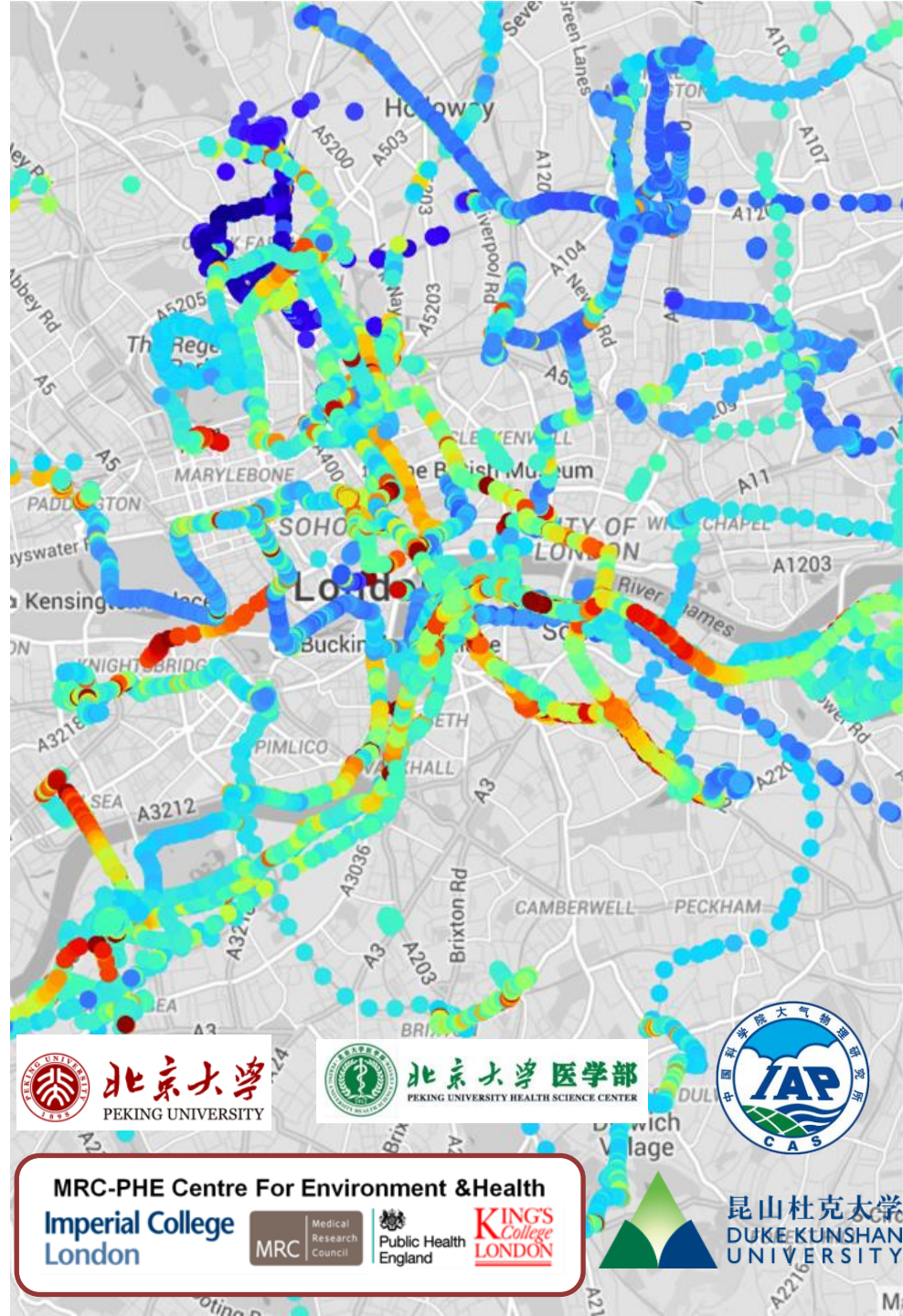


Quantifying personal exposure to air pollution with sensor technologies and digital science

Dr Lia Chatzidiakou
ec571@cam.ac.uk
the Rod Jones' group

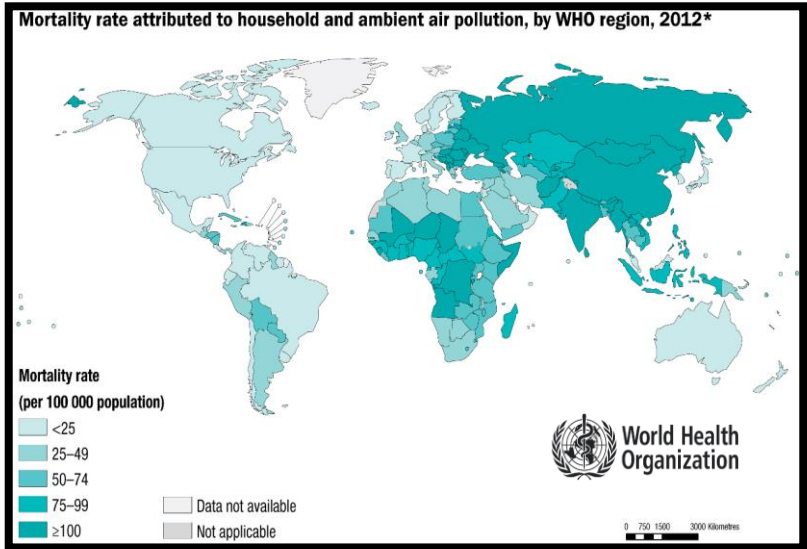
Yusuf Hamied Department of Chemistry
University of Cambridge

22nd November 2023

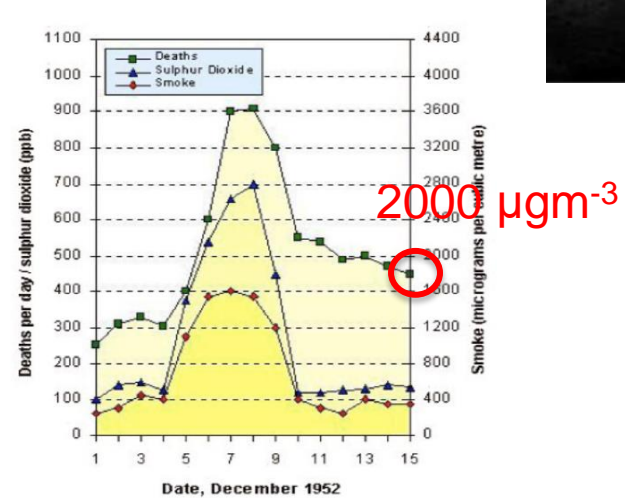


Air pollution and human health: a national problem but a global crisis

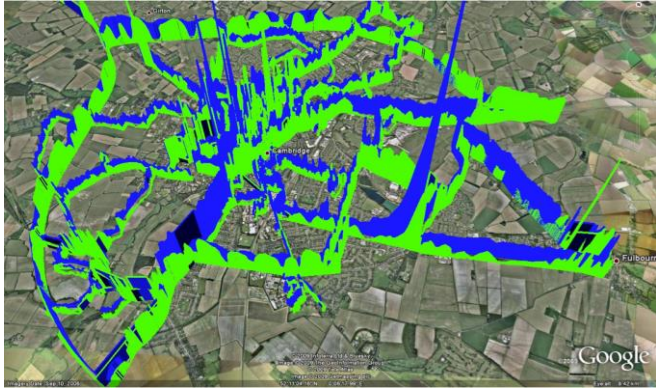
Estimated **8M premature deaths** annually globally



Low-income countries
are **worst affected**
(92% of pollution-related deaths)



Common air pollutants (reminder)



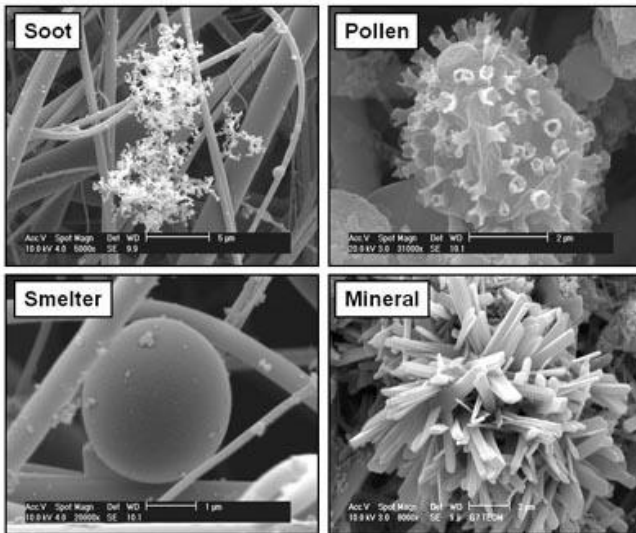
NO and NO₂ levels in Cambridge on one afternoon

NO_x (NO + NO₂)

- Primarily emitted by cars and trucks
- Associated with asthma attacks, respiratory illness and cardiovascular effects

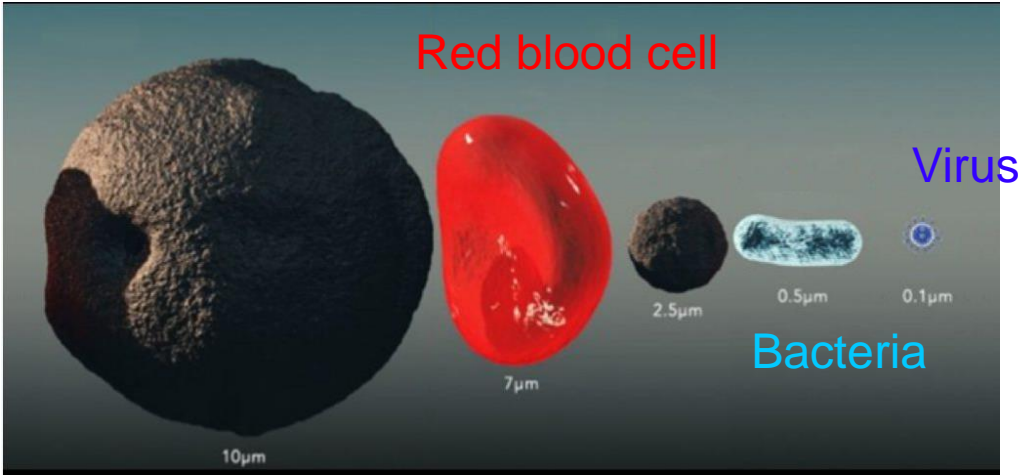
Particulate Matter (PM)

- Emitted directly or formed in the atmosphere
- Particles less than 2.5 micrometers in diameter, (fine particles or PM_{2.5}), pose the greatest risk to health.



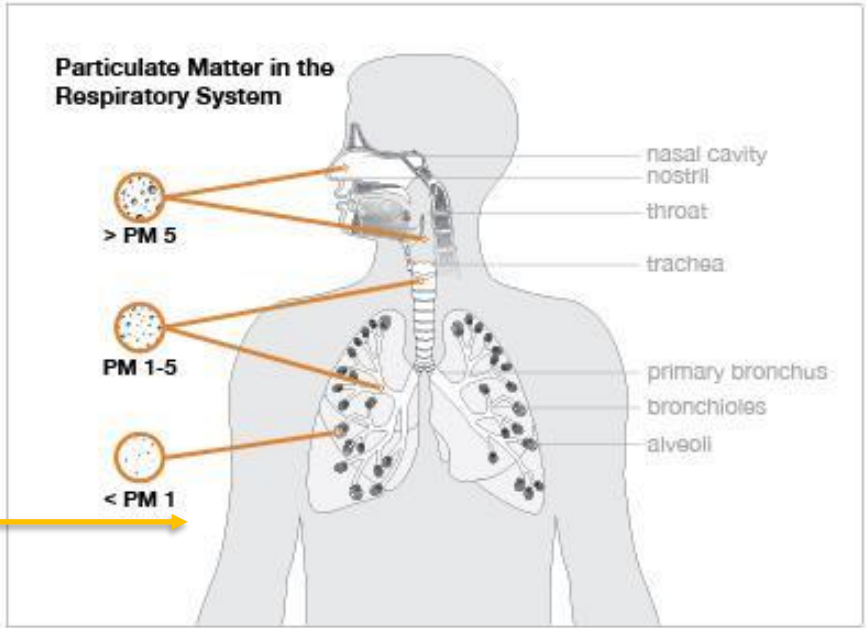
PM collected on filters and examined with a microscope. PM can be made up of different chemicals.

Particulate Matter (PM)



Particle size is important:

- Affects transport
- Time suspended in the air
- Related to chemical composition
- Deposition in the lungs



Measurement techniques for air quality

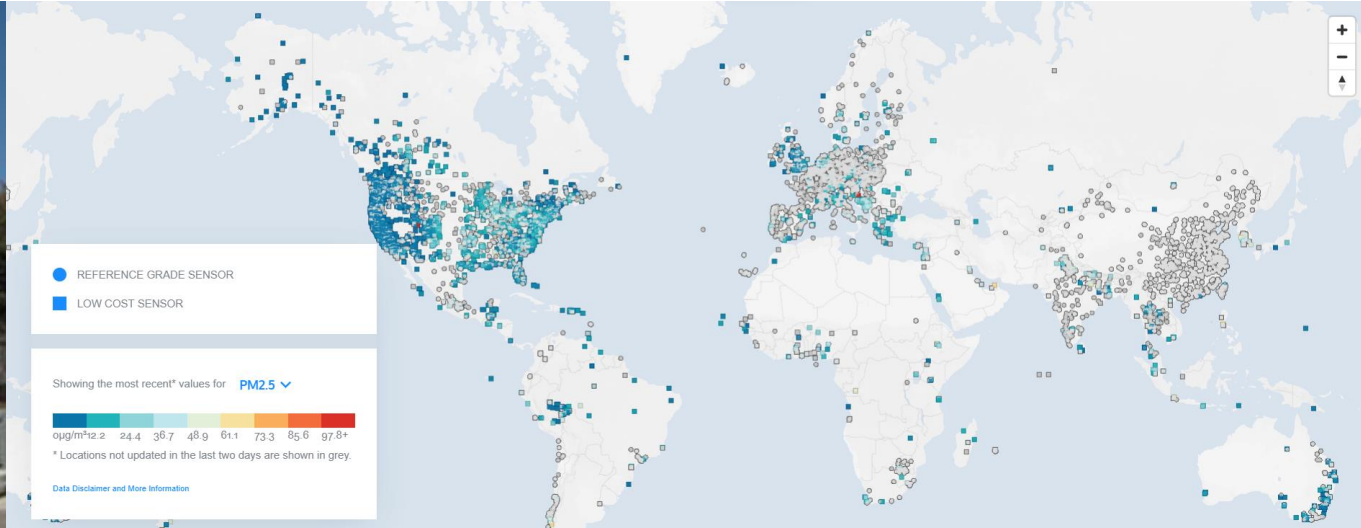
Reference-grade instrumentation



- **Expensive** to set up and maintain
- Require roadside **infrastructure** to house them
- Well-quantified **accuracy**

Measurement techniques for air quality

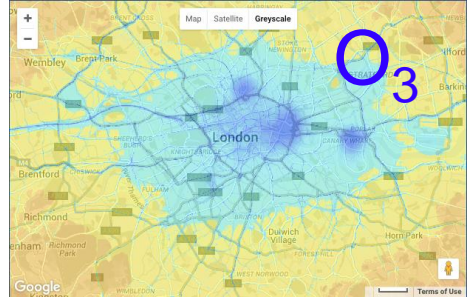
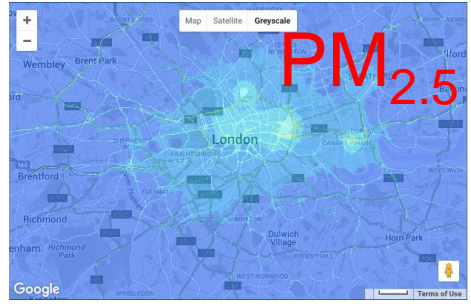
Reference-grade instrumentation



- **Expensive** to set up and maintain
- Require roadside **infrastructure** to house them
- **Well-quantified accuracy**

⇒ **Less resources where most needed**

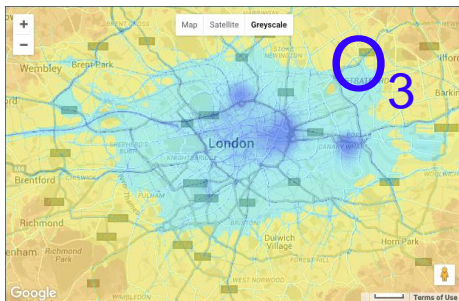
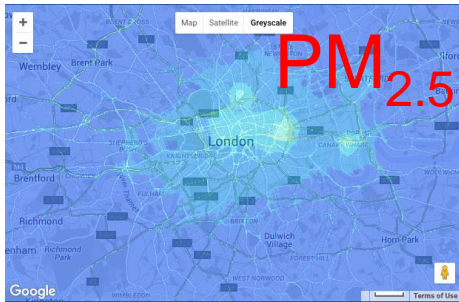
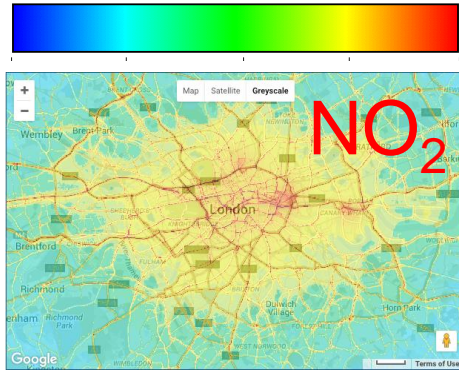
Air pollution and human health: what do we know?



Modelled annual averages

Air pollution and human health:

what do we know?

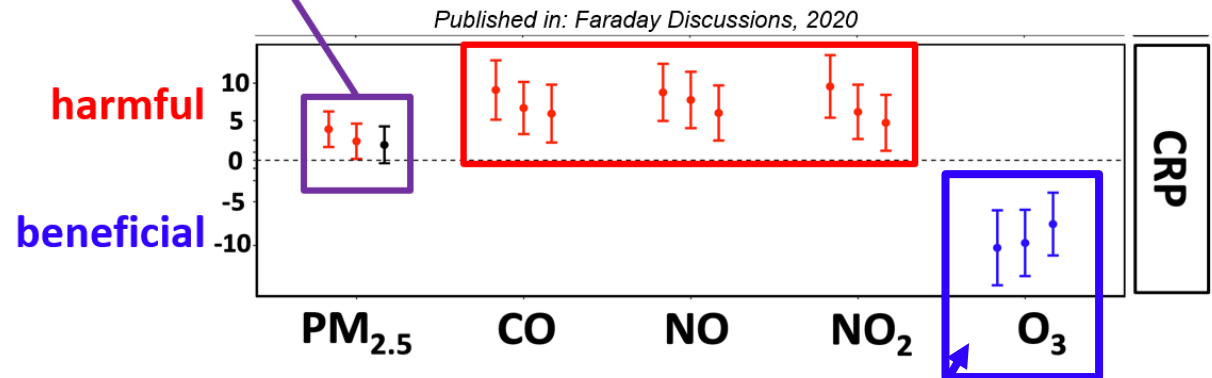


Modelled annual averages

- Single and dual pollutant models for NO₂ and PM_{2.5} all only explain the *same fraction* of health outcomes..... (UK COMEAP)

Lags of 1,2,3 days

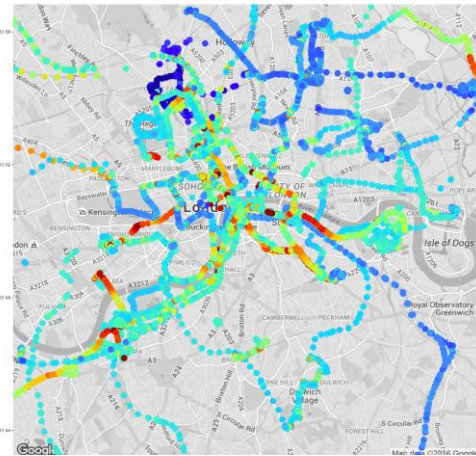
All these show the same effect - why?



- Some studies suggest surface O₃ can have a *protective effect*? ...O₃ often anti-correlated with other pollutants

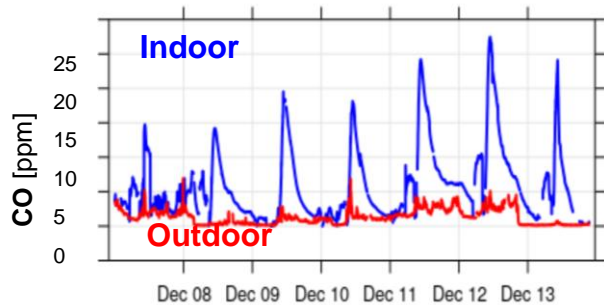
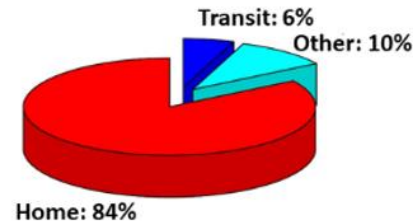
⇒ Cannot reliably distinguish causal links

Critical knowledge gaps: personal and indoor exposure



Map of personal exposure to NOx during commuting (London)

- Activity patterns
- Time budget

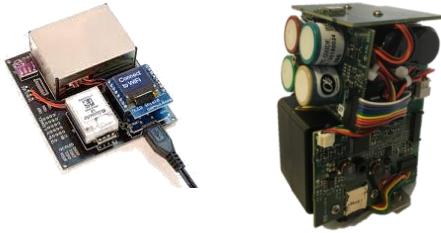


- Indoor air quality (**indoor sources**, vs **outdoor air**) very different from ambient pollution

⇒ **Linking activity to exposure and to health**

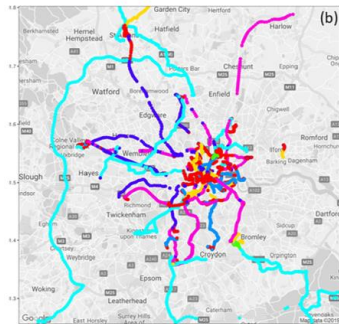
State-of-the-art health models: the next step

1. Sensor networks



Published in:
Chatzidiakou et al., Atmos. Meas. Tech., 2019

3. Automated time - activity patterns



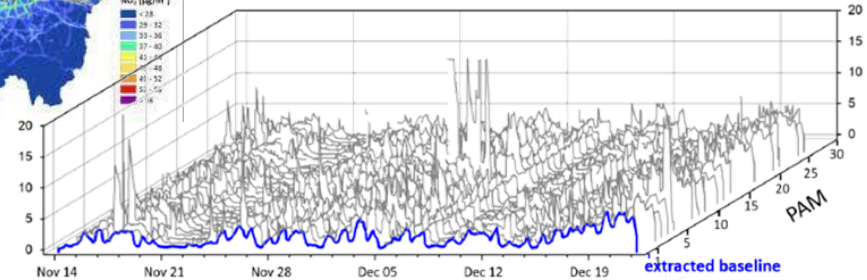
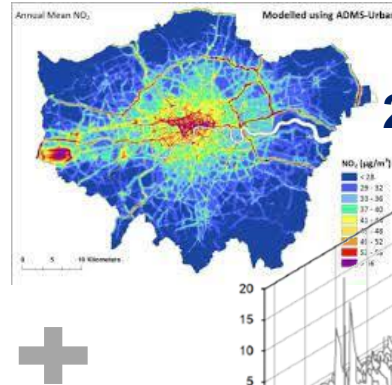
Mode of commuting

- walking
- cycling
- running
- car
- bus
- tube
- train



Published in:
Chatzidiakou et al., BMC Env. Health, 2022

2. Data assimilation methods



4. Novel health and exposure metrics

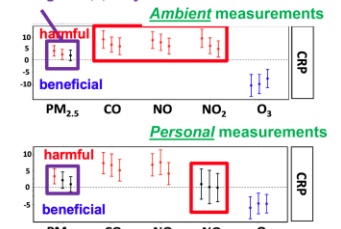
Ambient
CO, NO, NO₂ all show the same (harmful) effect - why?

O₃ can have a protective effect? ...O₃ often *anti-correlated* with other pollutants

PM_{2.5} harmful...

Personal
NO₂ no longer statistically significant.....

Relative (health) risks of different pollutants
Lags of 1,2,3 days



Published in: Faraday Discussions, 2020

- “Everyone’s disease is the product of the individual history of exposures, superimposed on their underlying genetic susceptibilities”
- Beyond hard and rare outcomes...

Sensor networks

Observational technologies

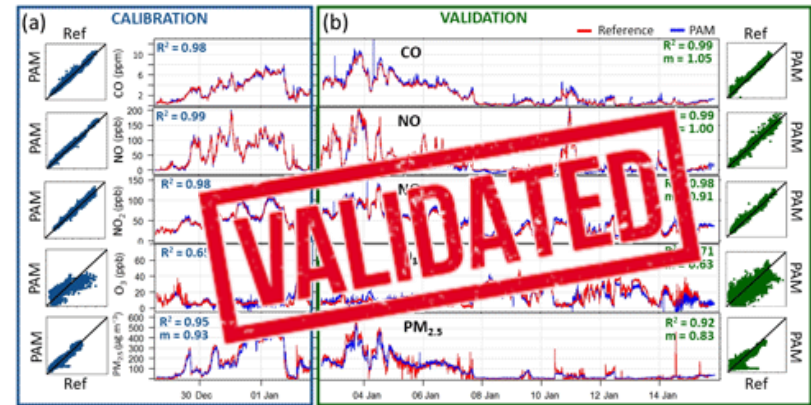


Exposure

- O_3 , NO_2 , NO , CO
- Size-specified PM
- Noise
- T, RH

Activity

- GPS
- Accelerometer
- Noise



⇒ Integration of new and historic data from multiple sources, over a variety of scales, resolutions and frequencies

Published in:
Chatzidiakou et al., AMT, 2019

Observational technologies



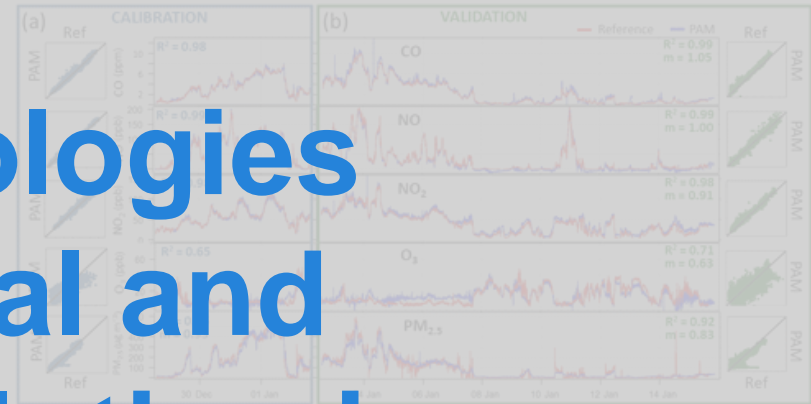
Sensor technologies increase spatial and temporal resolution in a way not possible before

Exposure

- O_3 , NO_2 , NO , CO
- SO_2 , H_2S , PM_{10}
- Noise
- T, RH

Activity

- GPS
- Accelerometer
- Noise



⇒ Integration of new and historic data from multiple sources, over a variety of scales, resolutions and frequencies

Published in:

Chatzidiakou et al., AMT, 2019

Data assimilation methods (indoor air)

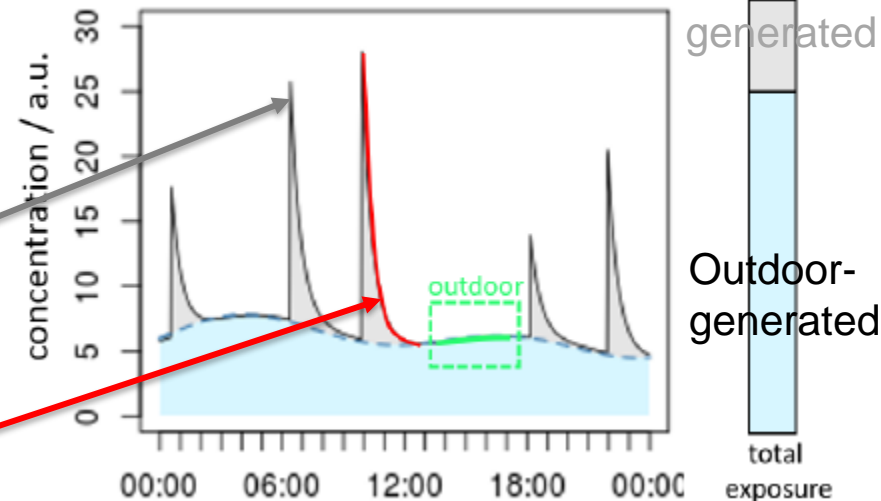
A simple model to estimate indoor air pollution levels

Indoor sources
(cooking, cleaning,
smoking etc)

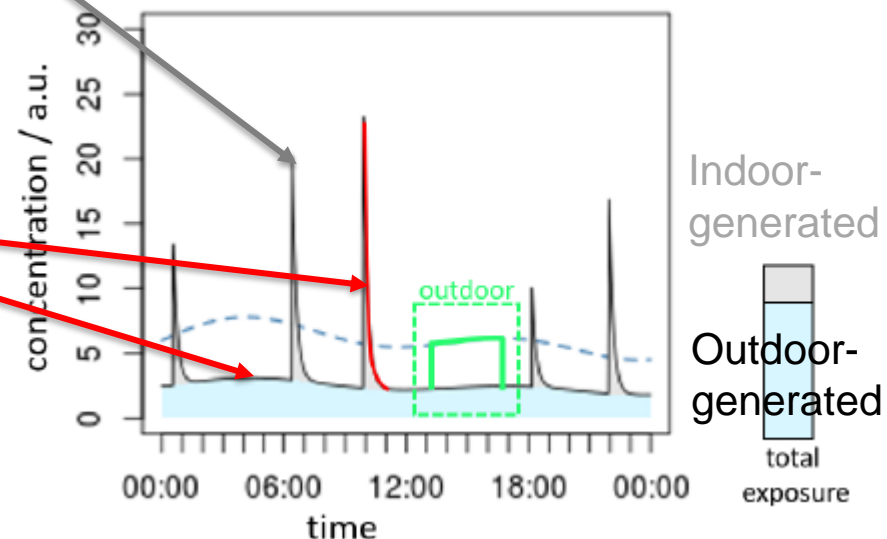
Ventilation to outdoors

Ventilation and other
processes (indoor
sinks)

(A) Inert pollutant



(B) Reactive pollutant



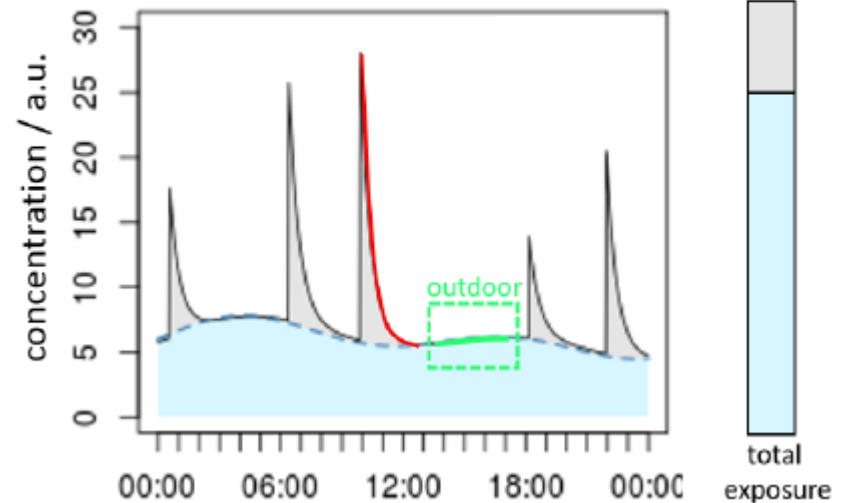
Data assimilation methods (indoor air)

Location, building characteristics, materials, operation and maintenance \Rightarrow
Large variation between and within microenvironments

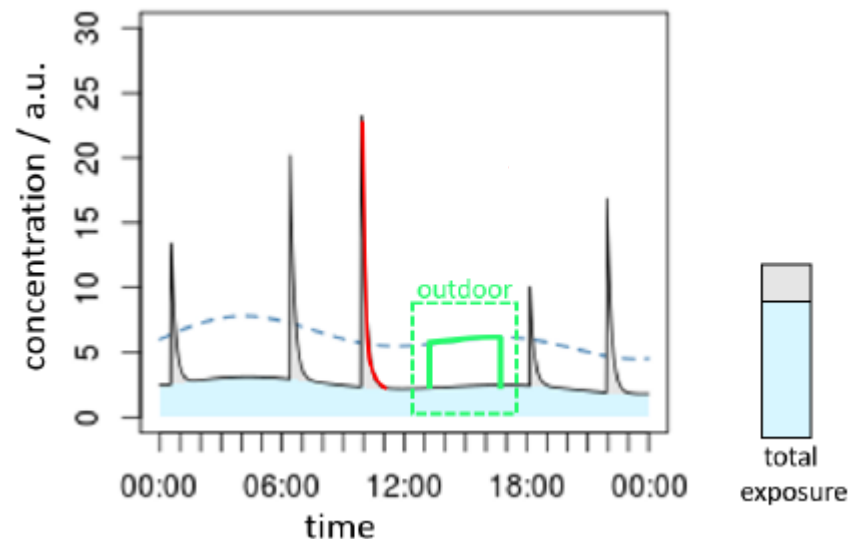
Examples of a “Victorian” and contemporary school with notable differences in indoor air pollution (London)



(A) Inert pollutant



(B) Reactive pollutant



Data assimilation methods (indoor air)

Location, building characteristics, materials, operation and maintenance \Rightarrow
Large variation between and within microenvironments

Examples of a “Victorian” and contemporary school with notable differences in indoor air pollution (London)



(A) Inert pollutant

The Guardian
Newspaper of the year

Air pollution worse inside London classrooms than outside, study finds

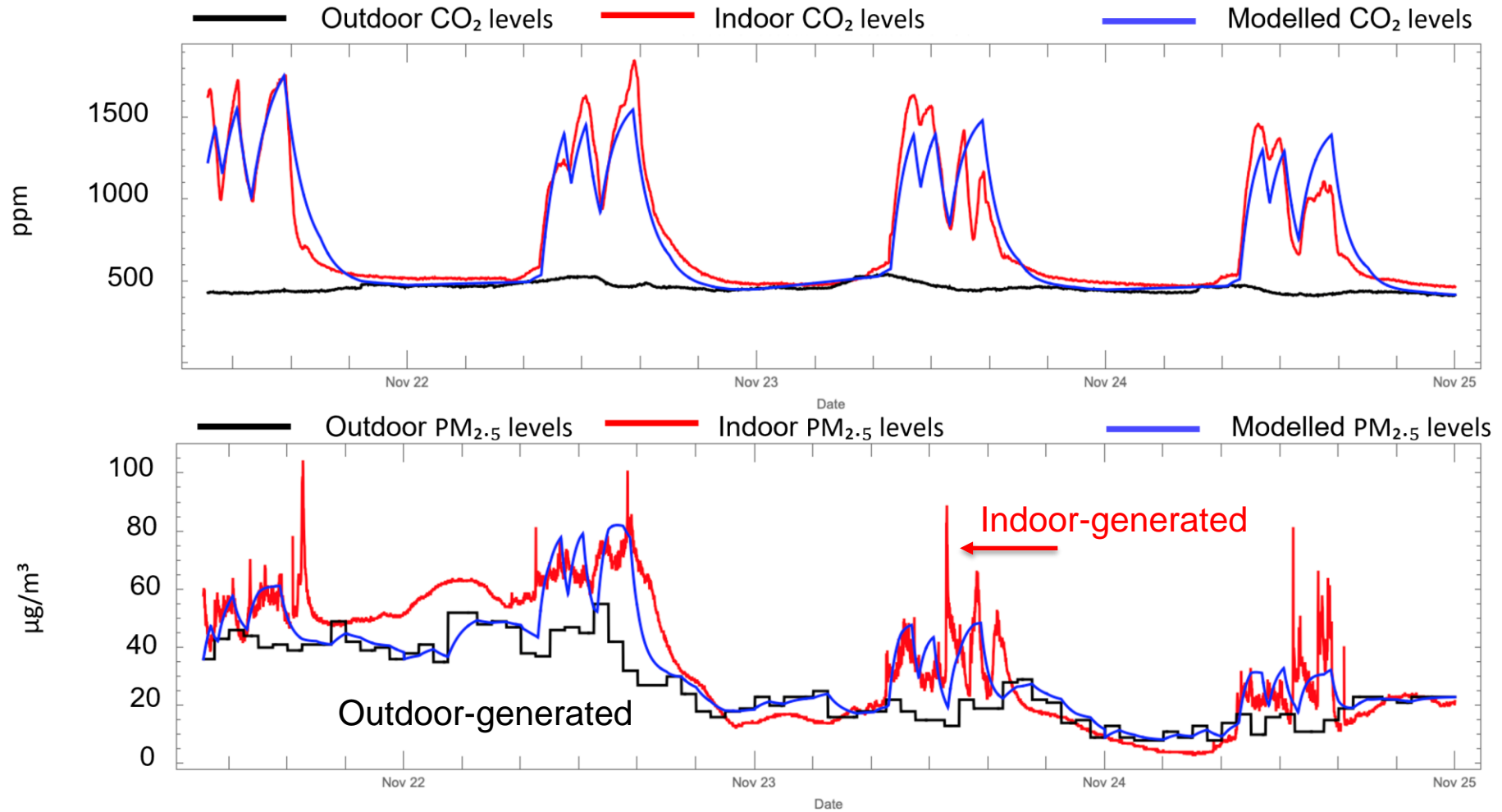
Exclusive: study of schools in capital finds dangerous levels of fine particulate pollution within classrooms, putting children at risk



📹 The mayor of London, Sadiq Khan, has announced a £1m fund to help tackle air pollution in schools. Photograph: Stefan Rousseau/PA Images

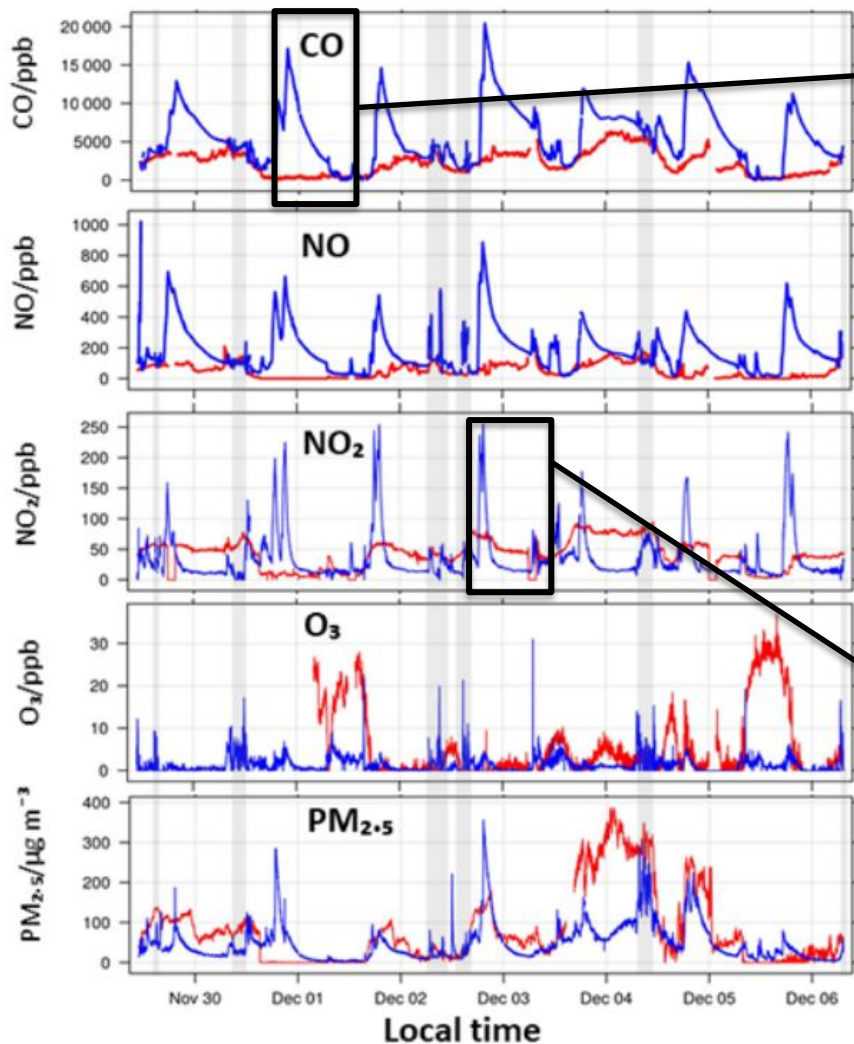
00:00 06:00 12:00 18:00 00:00
time total exposure

Outdoor measurements vs personal exposure (school)

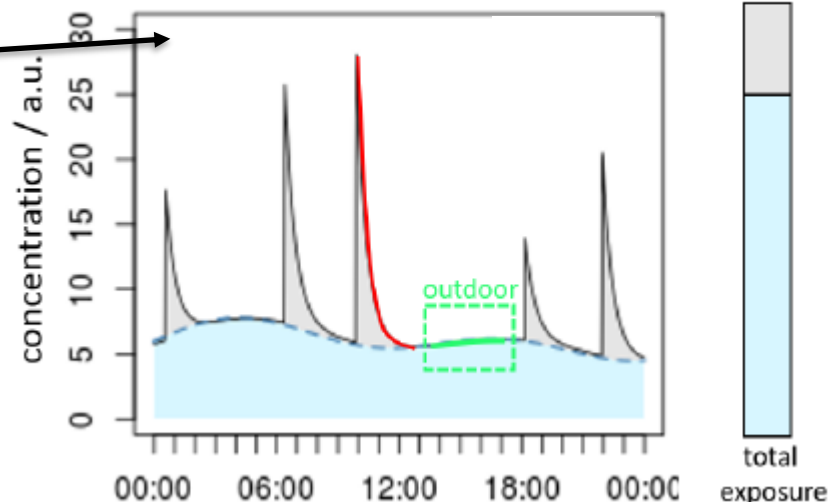


Outdoor measurements vs personal exposure (home)

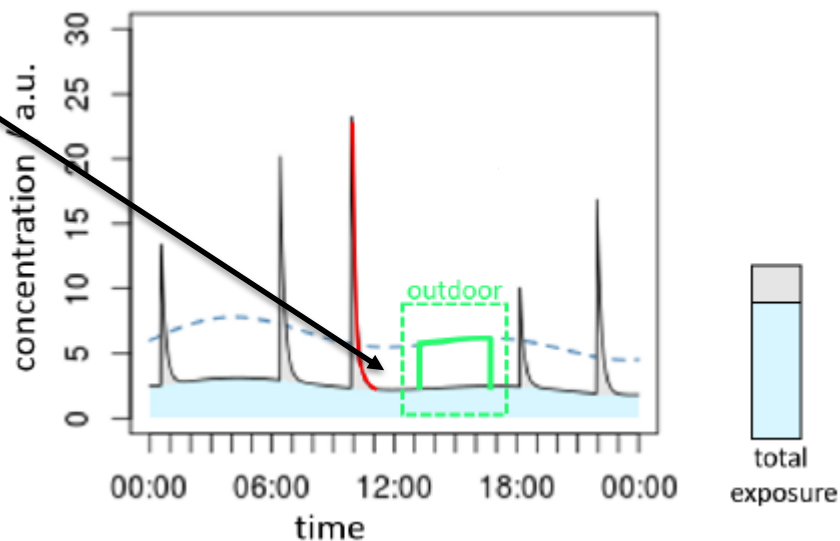
— Personal measurements — Monitoring station — Outdoors



(A) Inert pollutant

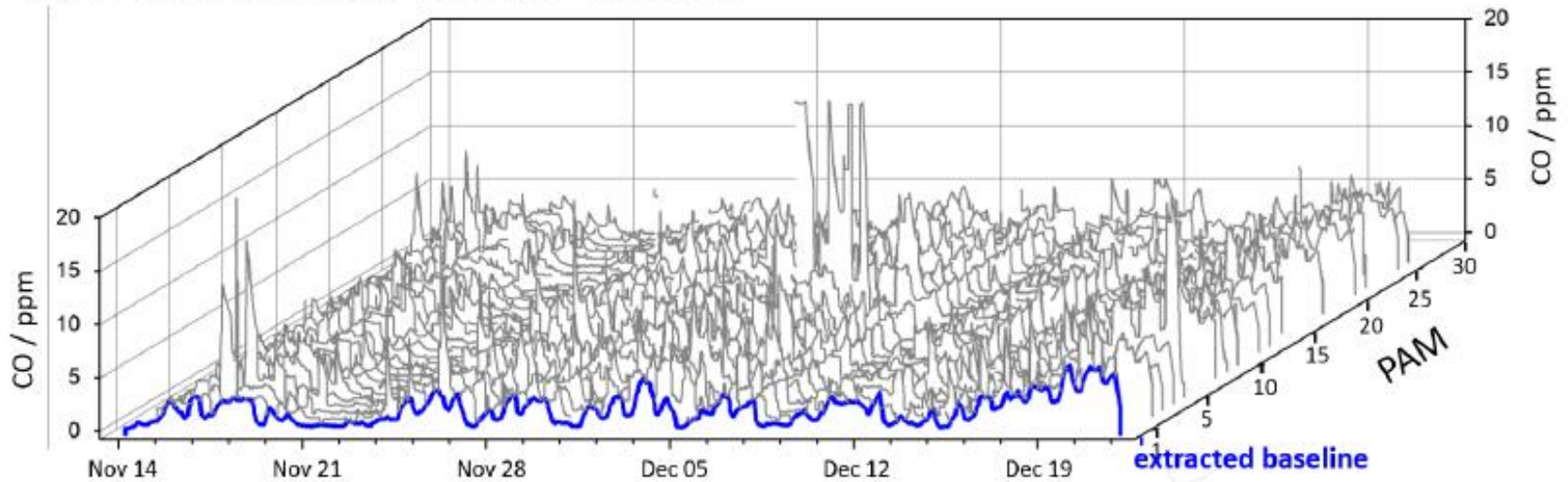


(B) Reactive pollutant

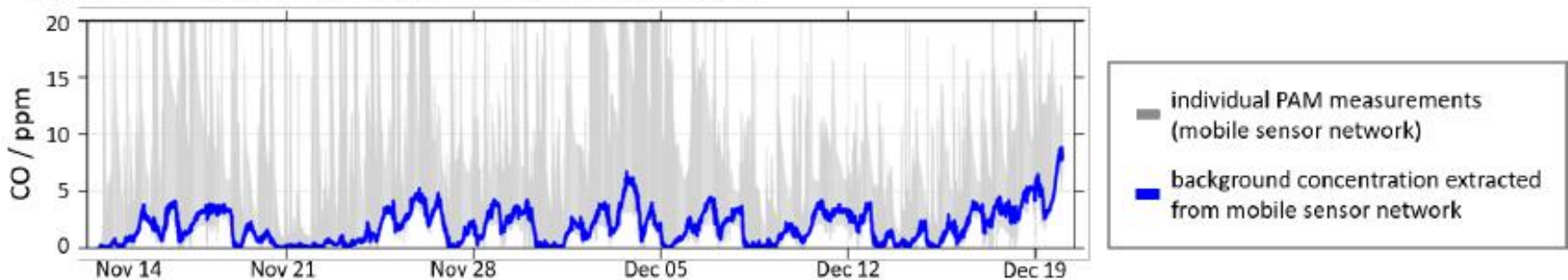


Outdoor measurements vs personal exposure (sensor network)

(a) 3D view, short term exposures removed

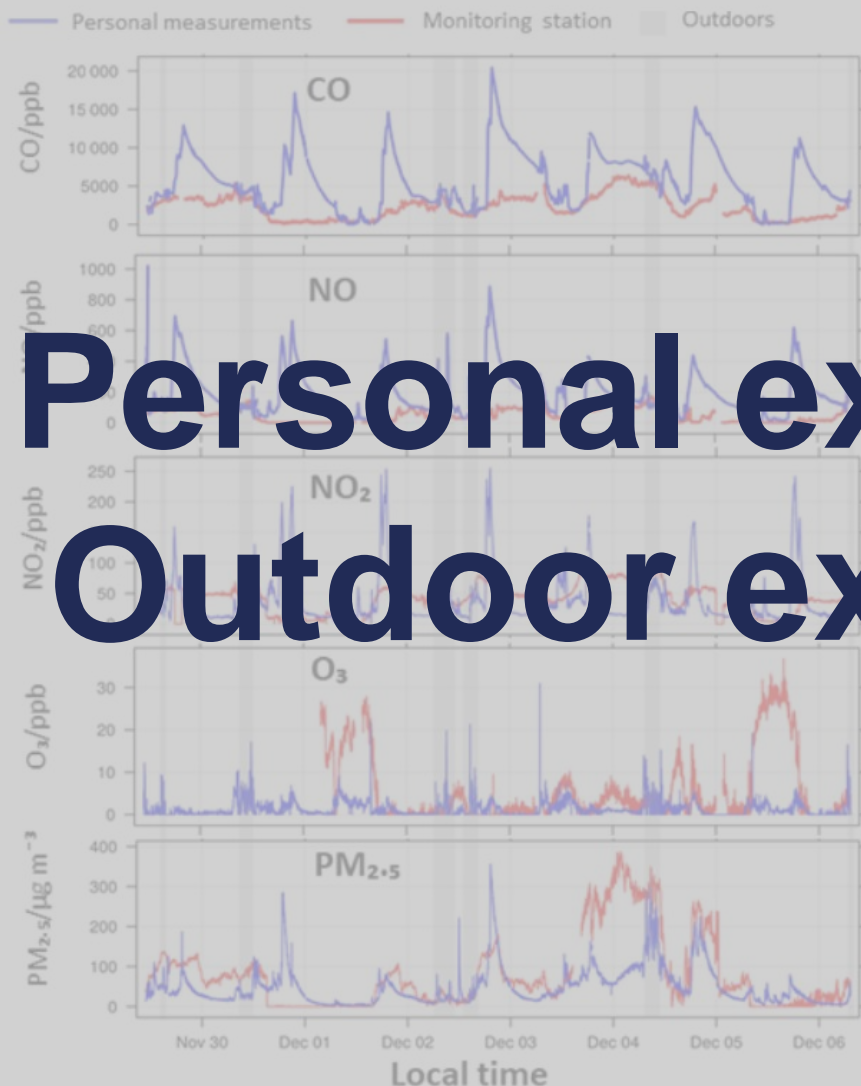


(b) frontal view, including short-time exposures

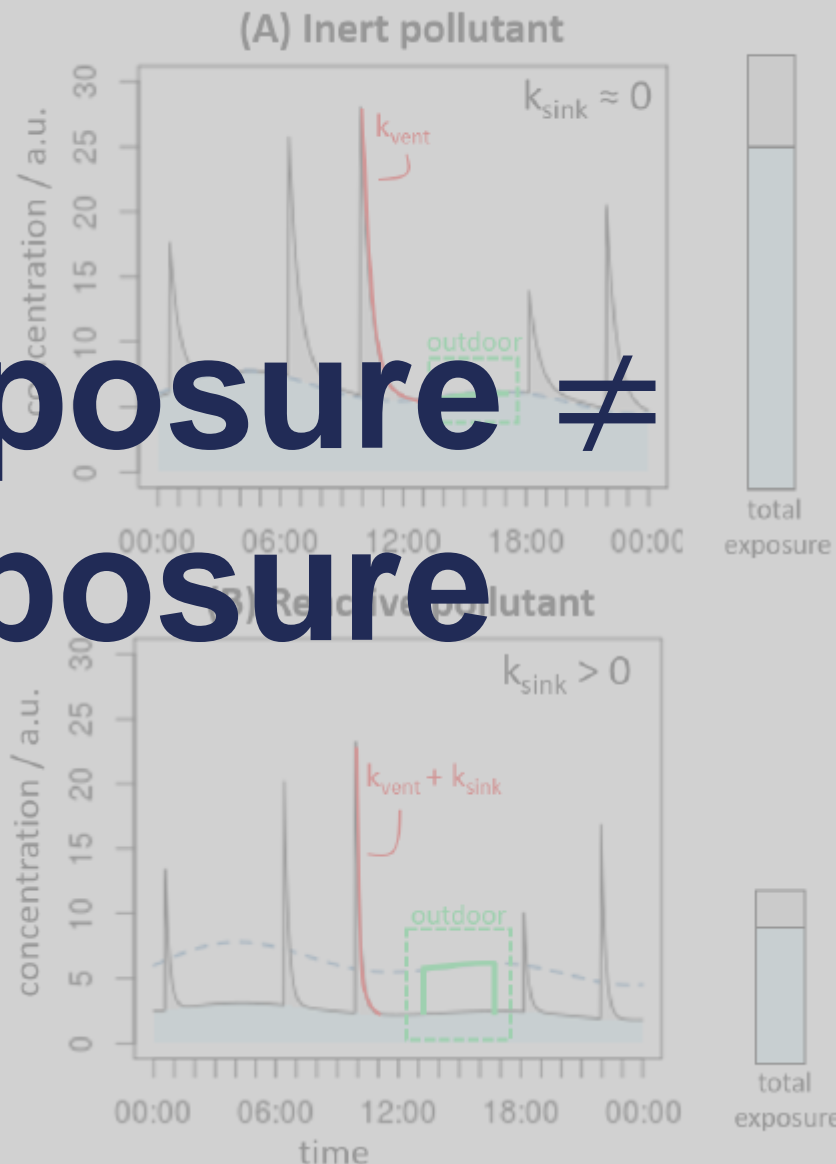


⇒ Separate the indoor-
from the outdoor-generated
component of exposure !

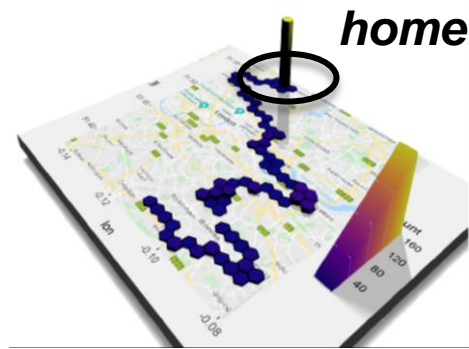
Are outdoor measurements good proxies of personal exposure?



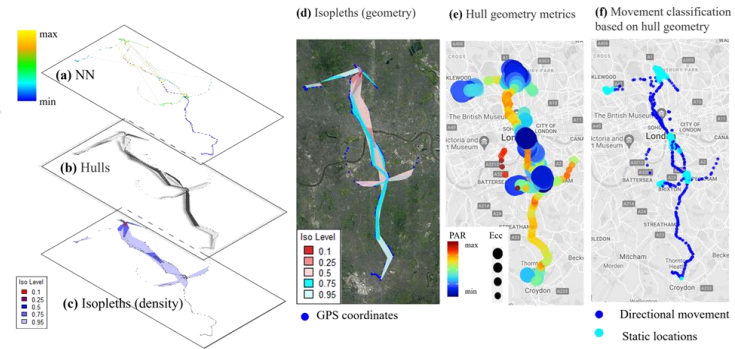
Personal exposure ≠ Outdoor exposure



Automated time-activity patterns



Geo-spatial
and temporal
clustering

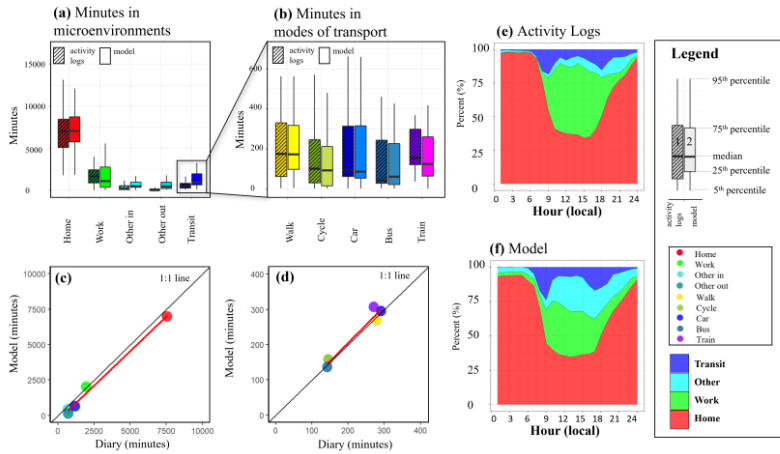


Mode of transport
classification
(AI algorithm)

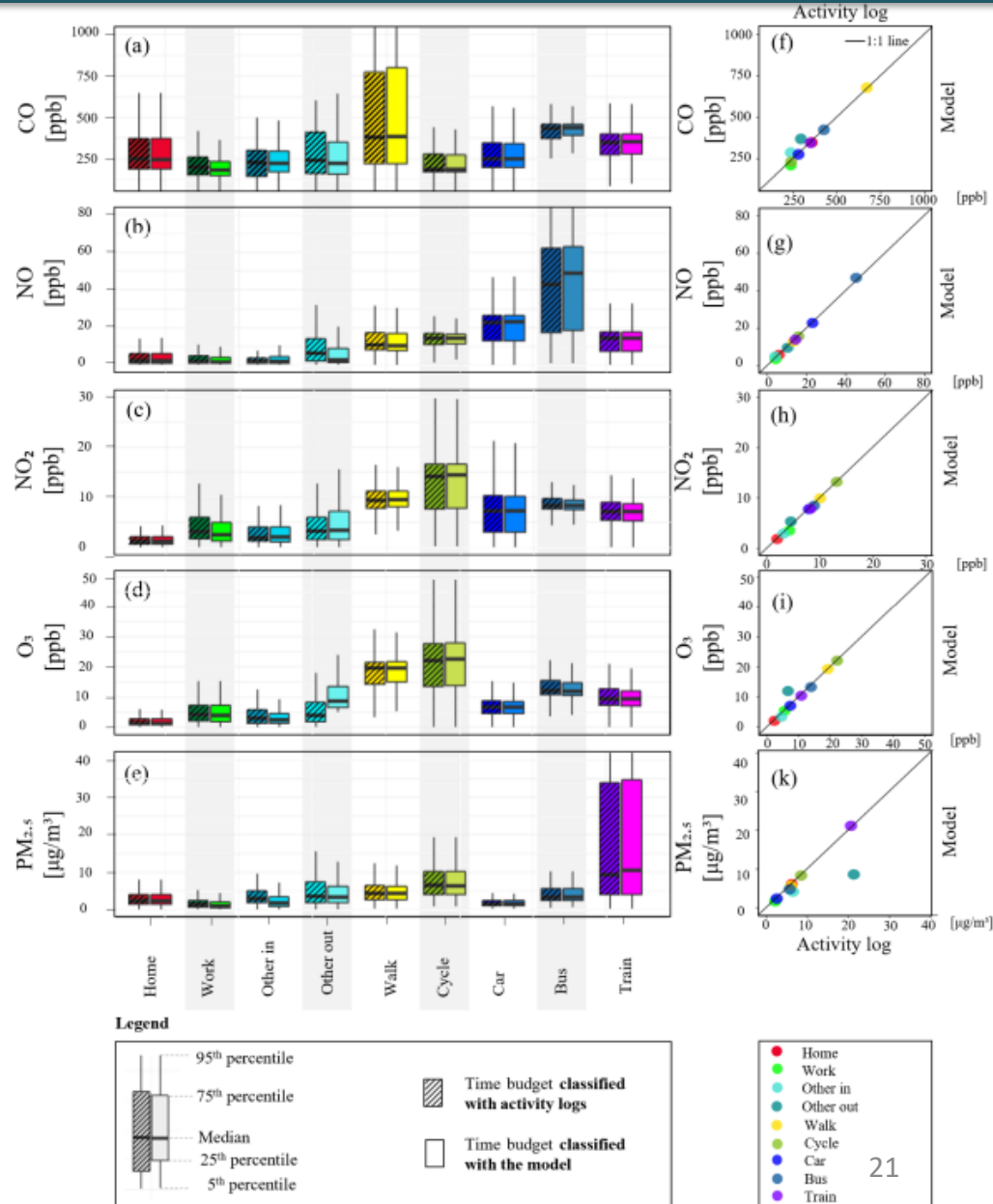


⇒ Automated interpretation with advanced spatial analysis, AI and innovative methods

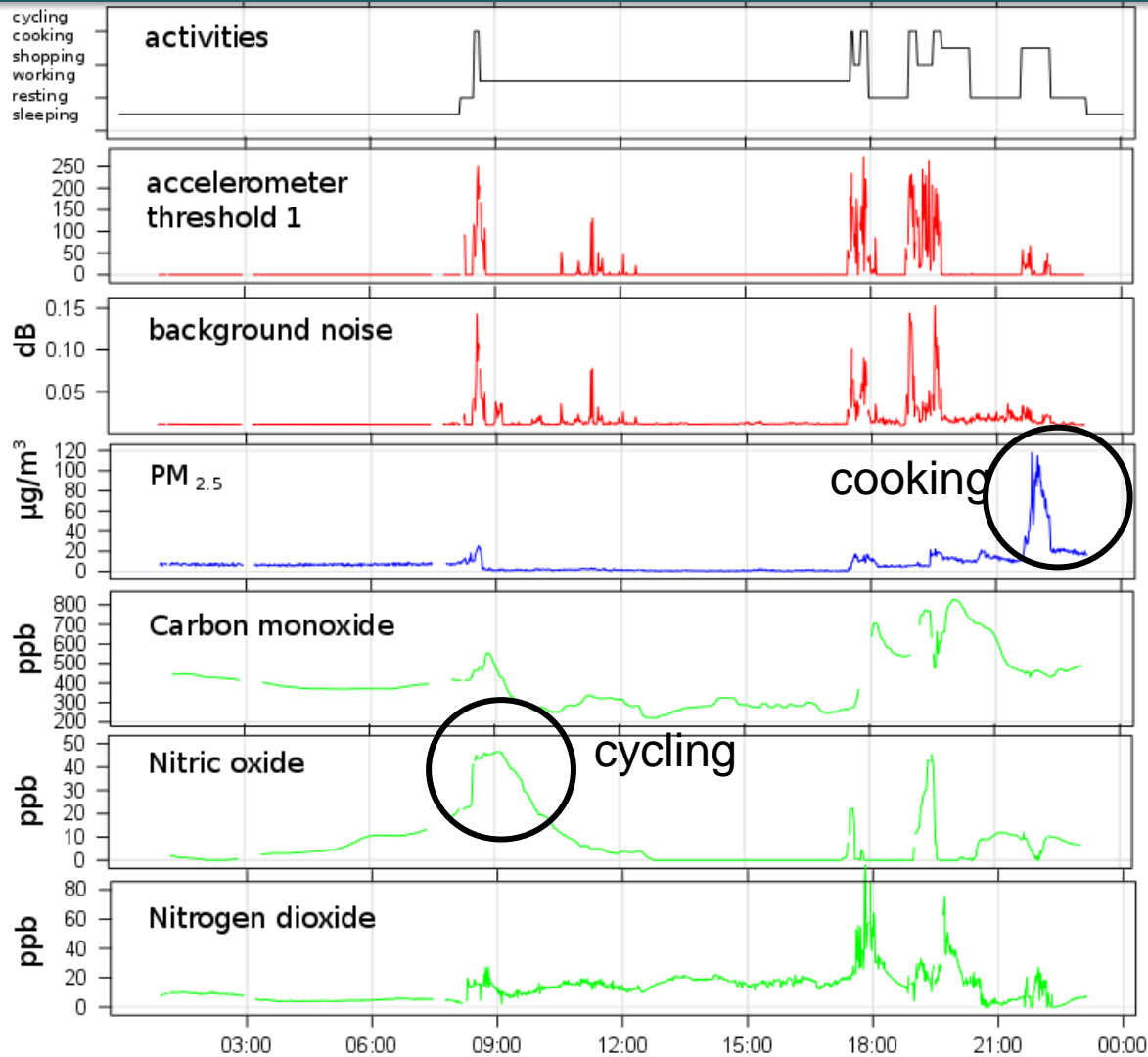
Automated time-activity patterns



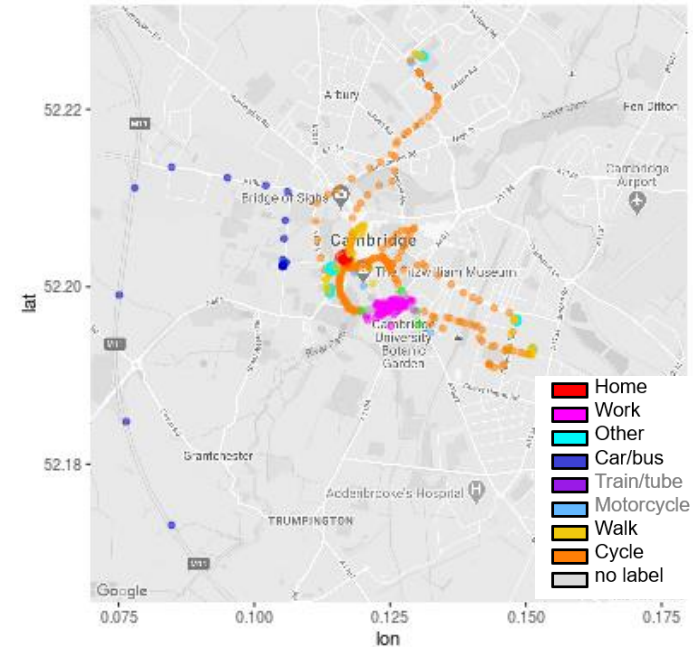
- ⇒ Commuting small fraction of time-budget
- ⇒ Maximum exposure to PM in the London Underground
- ⇒ Maximum exposure to NO_x and ozone during street-level commuting



Deployment of a PAM: illustrative example



Peak exposure events due to local outdoor (traffic) and indoor (cooking) sources cannot be captured with outdoor networks

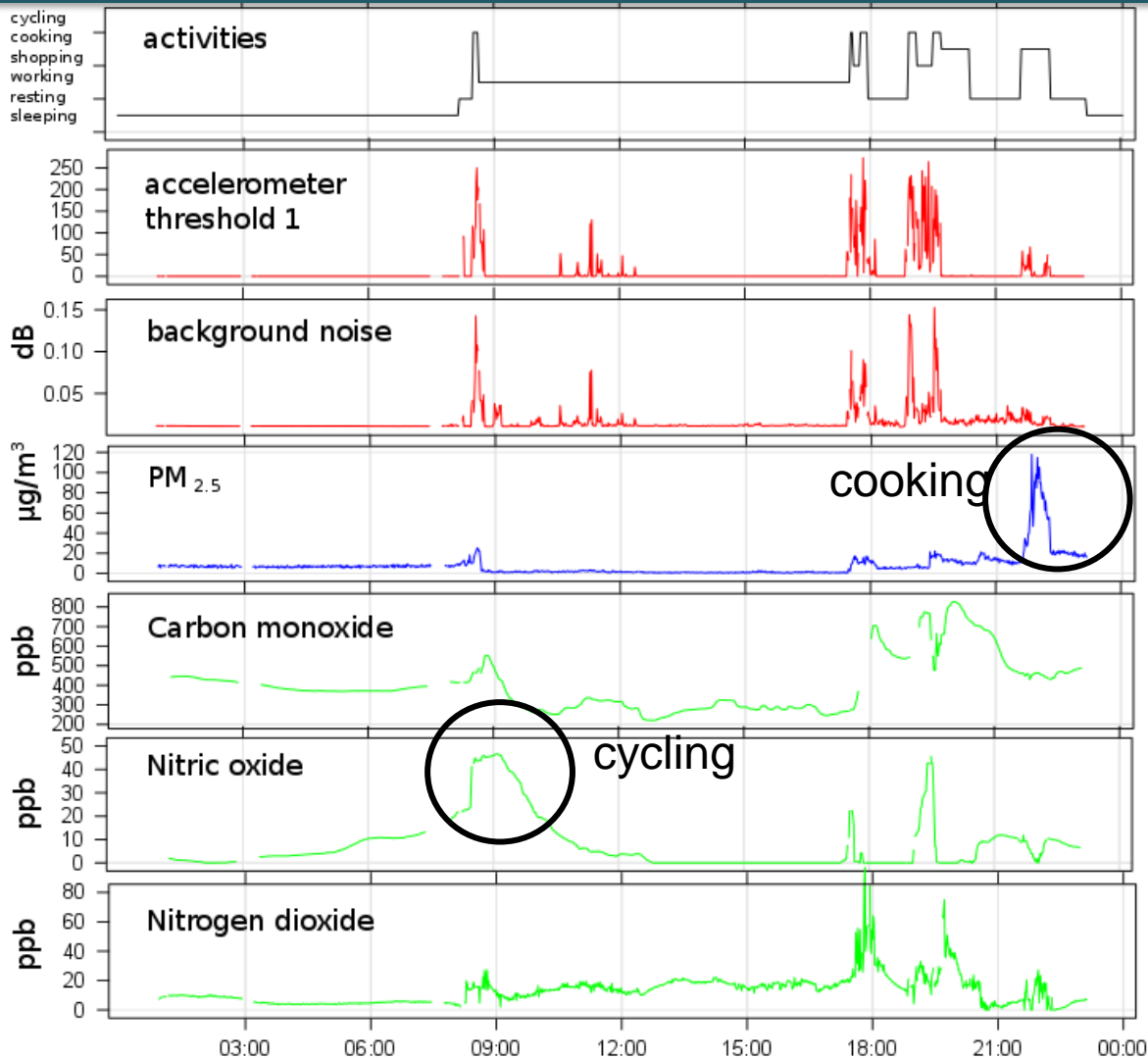


Time series of PAM parameters during a typical day

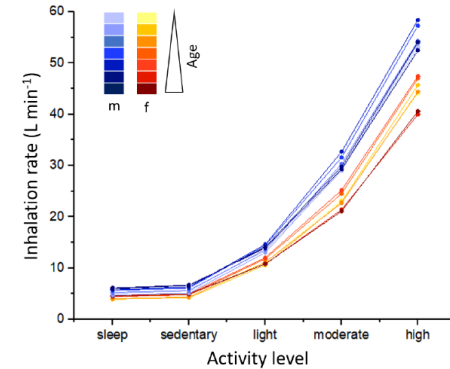
Visited locations and commuting mode during one typical week



Why physical activity important?

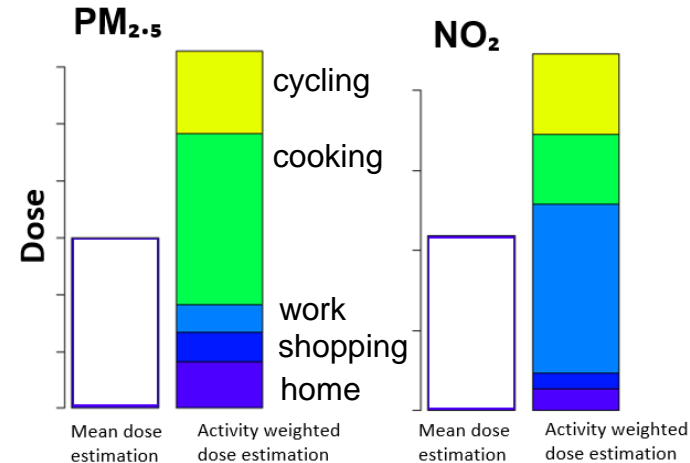


Time series of PAM parameters during a typical day



Physical activity important for dose!!!

Total exposure * inhalation rates



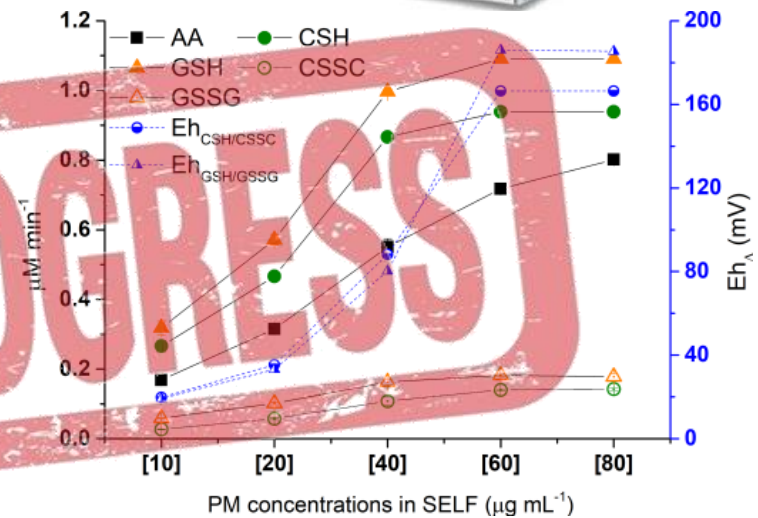
Why physical activity important?

- ⇒ Different sources operate in different microenvironments
- ⇒ Affect chemical composition
- ⇒ Particle toxicity is very different !!!

The *oxidative potential* of inhalable airborne particles is a measure of their ability to directly cause oxidative stress in the lung by depleting the antioxidants naturally present in the lung fluids.

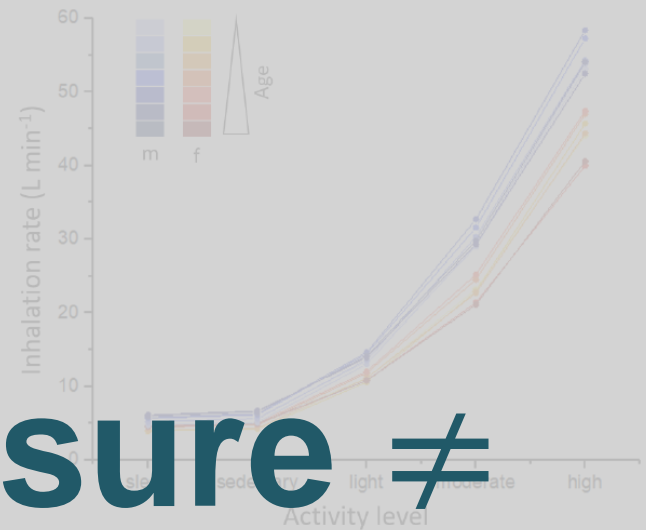
- Extraction of inhalable and respirable PM samples in a surrogate lung fluid containing glutathione and ascorbic acid (natural antioxidant).
- HPLC-MS analysis of the extracts.
- Measurements of the kinetics of the depletion of the natural antioxidants.

LTQ-Orbitrap Mass Spectrometer @UCAM.



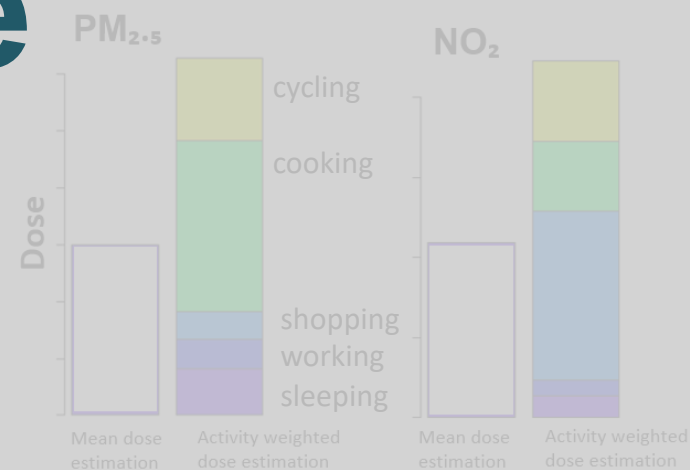
Example of depletion kinetics from Shahpoury et al. 2019, Atmos. Meas. Tech., 12, 6529–6539, 2019.

Deployment of a PAM: illustrative example



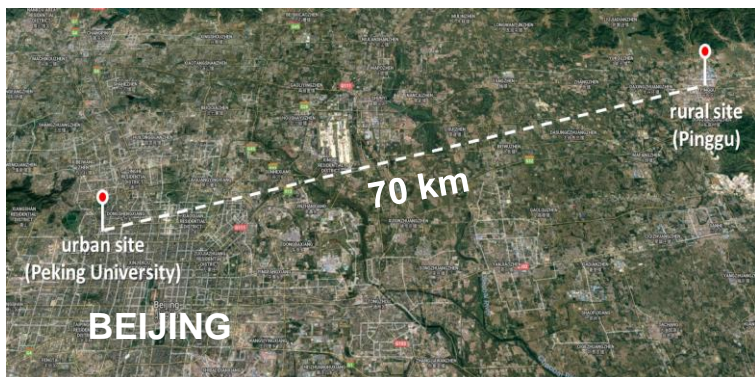
Personal exposure ≠ Personal dose

Total exposure * inhalation rates



Time series of PAM parameters during a typical day

The AIRLESS project



~ 250 participants

- 2 weeks summer/winter
- Parallel medical monitoring

Journal of Exposure Science & Environmental Epidemiology (2020) 30:981–989
<https://doi.org/10.1038/s41370-020-0259-6>

ARTICLE

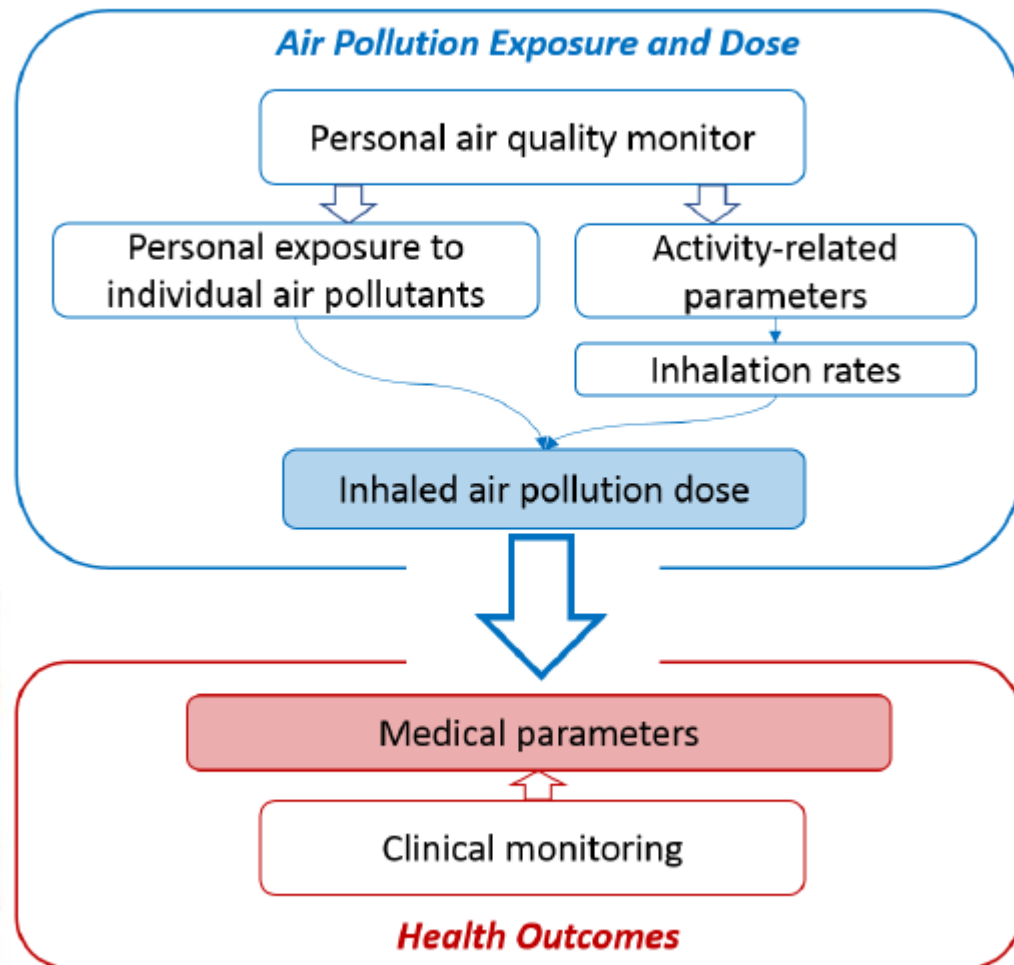
Using low-cost sensor technologies and advanced computational methods to improve dose estimations in health panel studies: results of the AIRLESS project

Lia Chatzidiakou¹ · Anika Krause¹ · Yiqun Han^{2,3,4} · Wu Chen¹ · Li Yan^{2,4} · Olalekan A. M. Popoc Mike Kellaway⁵ · Yangfeng Wu⁶ · Jing Liu⁷ · Min Hu^{3,8} · AIRLESS team · Ben Barratt^{2,4,9} · Frank J. I Tong Zhu^{3,4} · Roderic L. Jones¹

Received: 14 January 2020 / Accepted: 29 July 2020 / Published online: 12 August 2020
© The Author(s), under exclusive licence to Springer Nature America, Inc. 2020



Chatzidiakou et al., 2020



To establish more reliable links between air pollution exposure and health (in China)

Urban Beijing

High-rise residential blocks
Centralised heating



Peri-urban Beijing (Pinngu)

Agriculture activities
Biomass burning for domestic energy
(cooking, heating)



Atmos. Chem. Phys., 19, 7519–7546, 2019
<https://doi.org/10.5194/acp-19-7519-2019>
© Author(s) 2019. This work is distributed under
the Creative Commons Attribution 4.0 License.

Atmospheric
Chemistry
and Physics
EGU



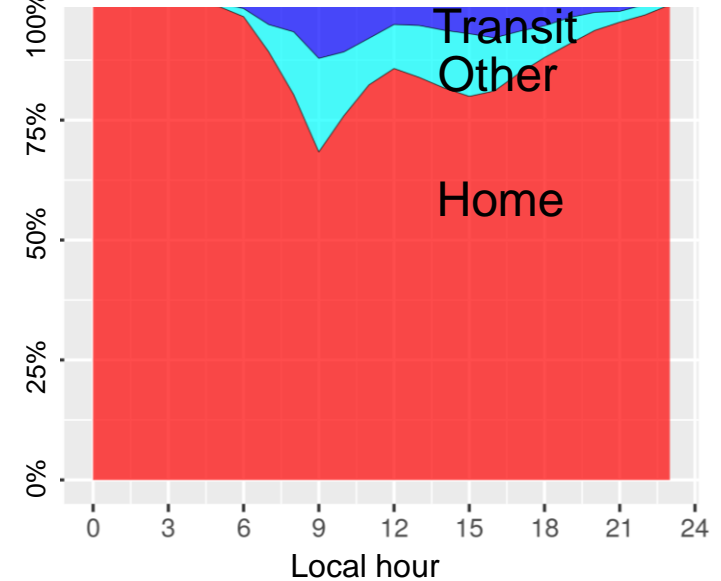
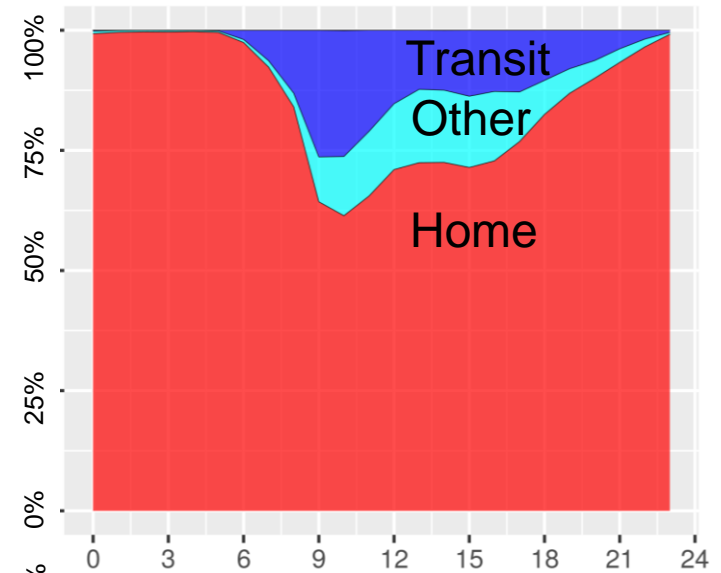
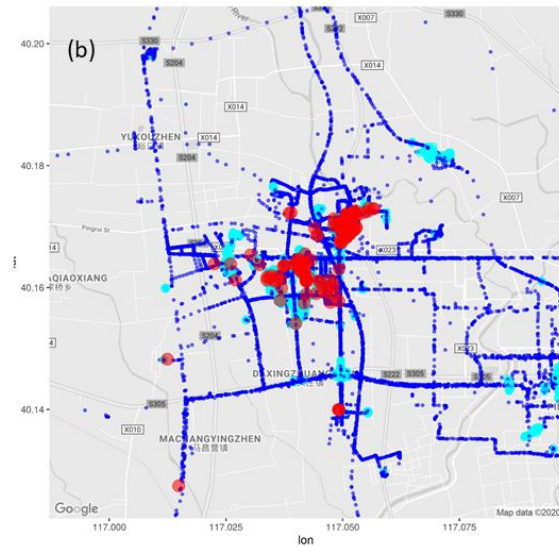
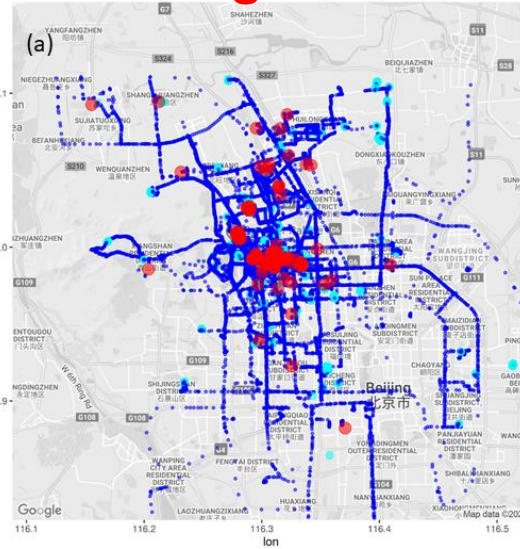
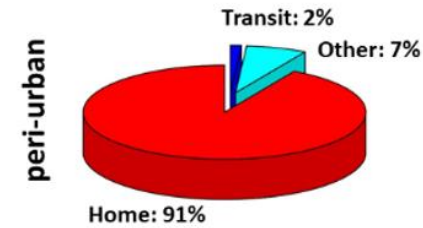
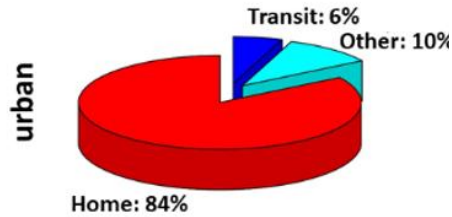
Personal activity vs ambient monitors...

1) *Over 80% of time spent indoors*

2) *Large spatial coverage*

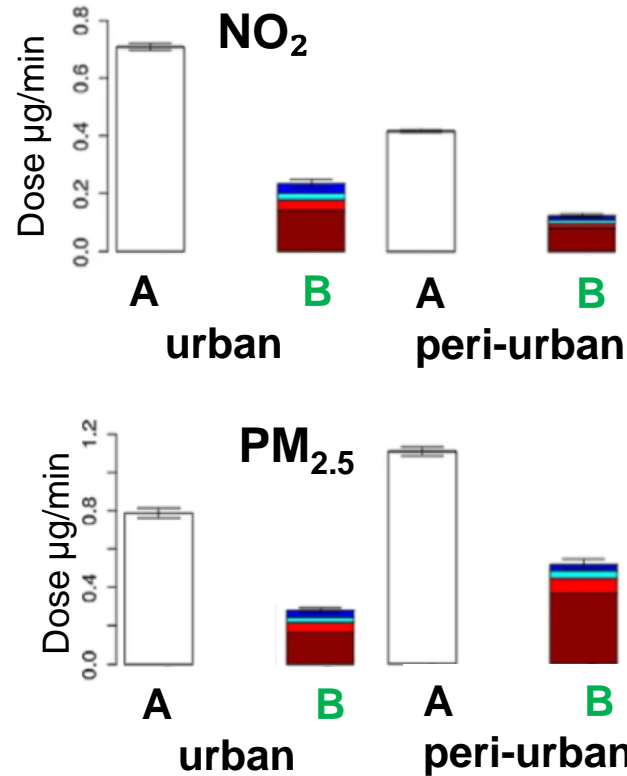
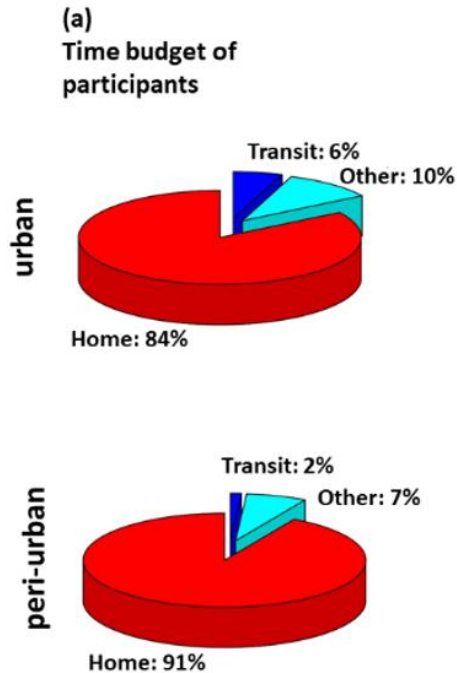
3) *Diurnal patterns*

(a) Time budget of participants



Personal activity vs ambient monitors...

1) *Over 80% of time spent indoors*



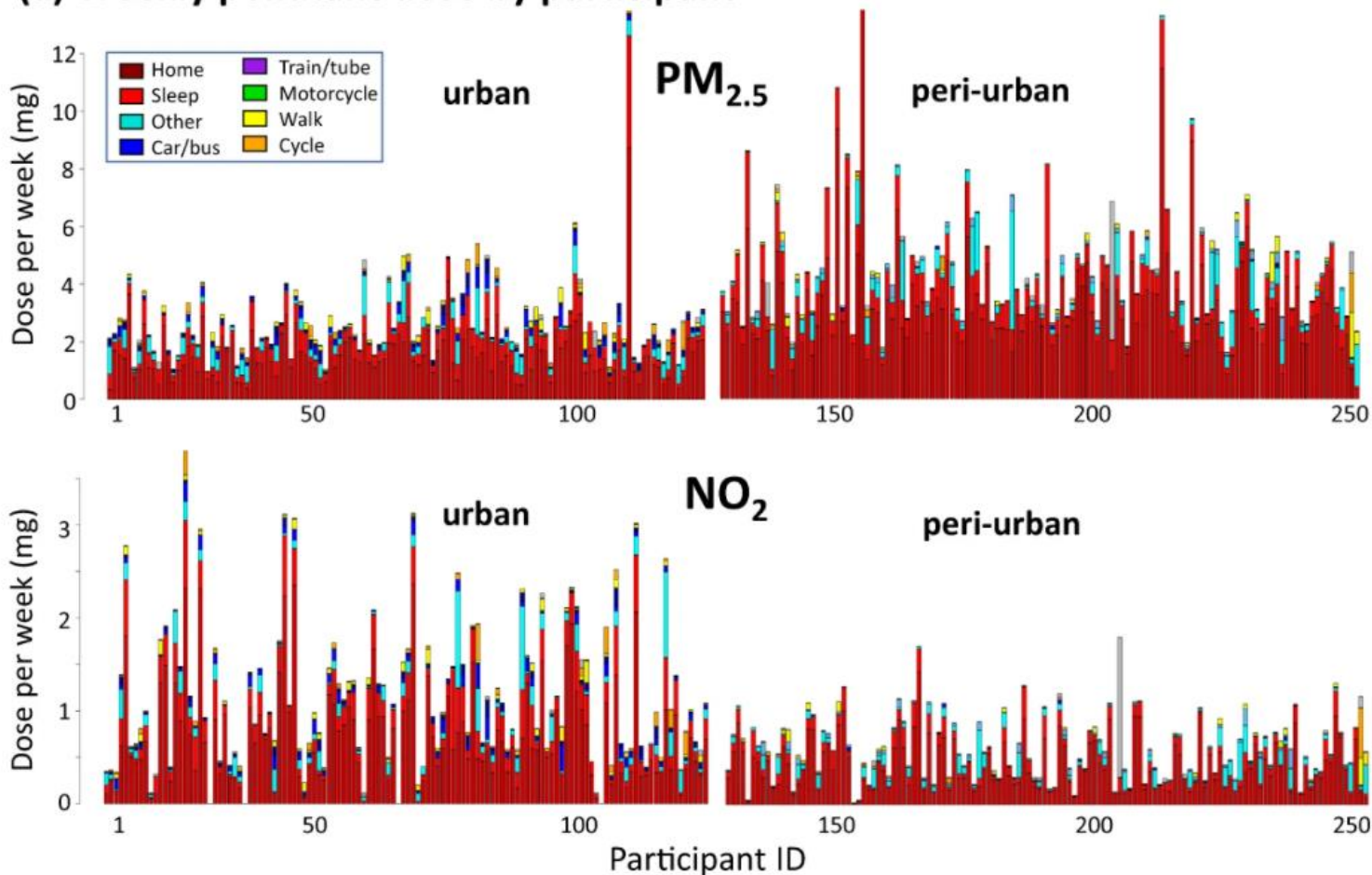
A: ambient measurements + generic inhalation

B: personal monitoring + activity

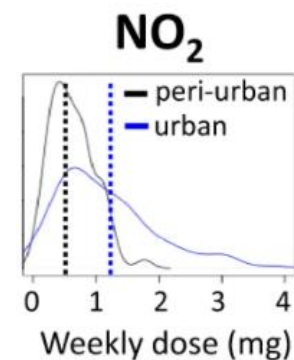
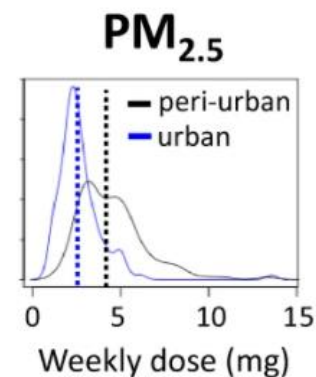
2) *Exposure not well represented (overestimated) by outdoor air quality.....*

The AIRLESS project: personal dose

(a) Weekly pollutant dose by participant



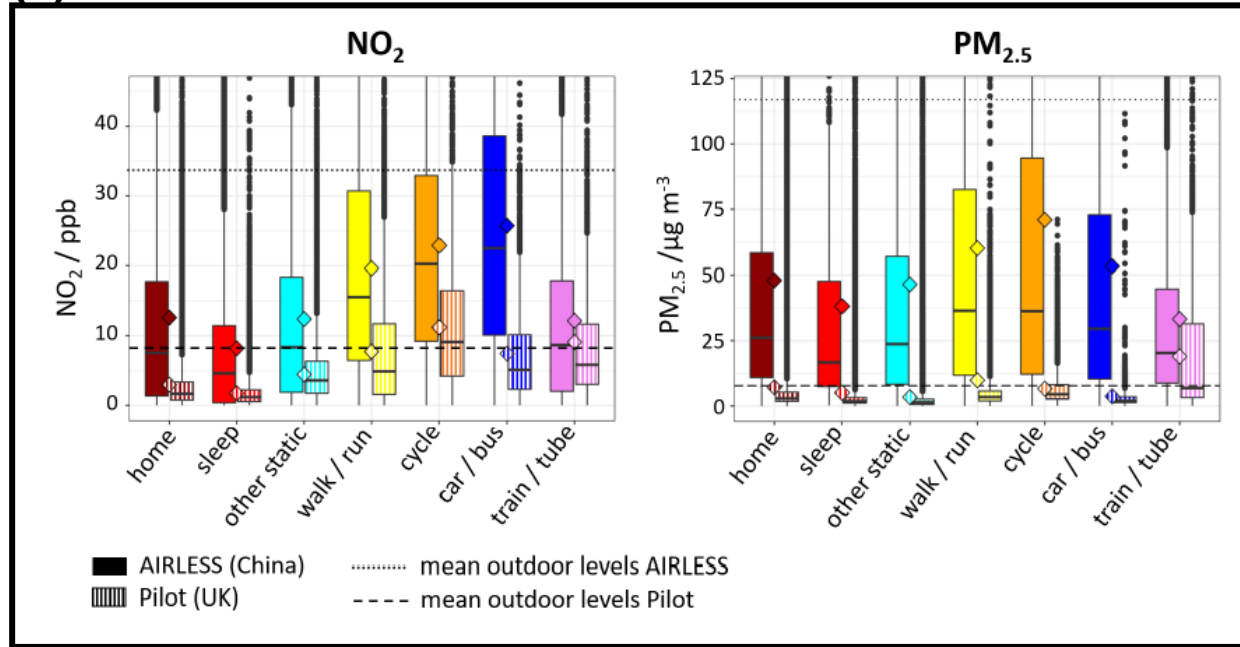
(b) Density plots



The home microenvironment was the most important modifier of personal dose

How does this compare to the UK?

(a) Absolute concentrations

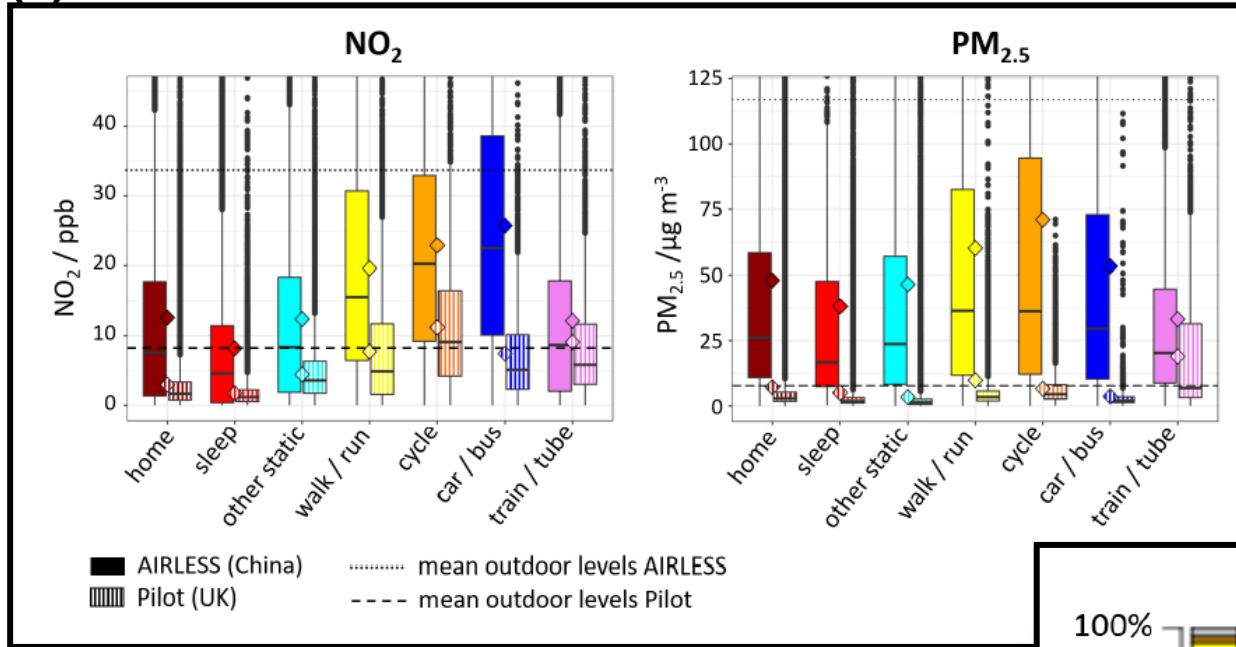


Pollutants monitored with PAMs in 250 AIRLESS participants and 35 London participants during a week.

- 1) **Outdoor affects personal exposure**
- 2) **CHINA: home exposure high**
- 3) **UK: PM in underground high**

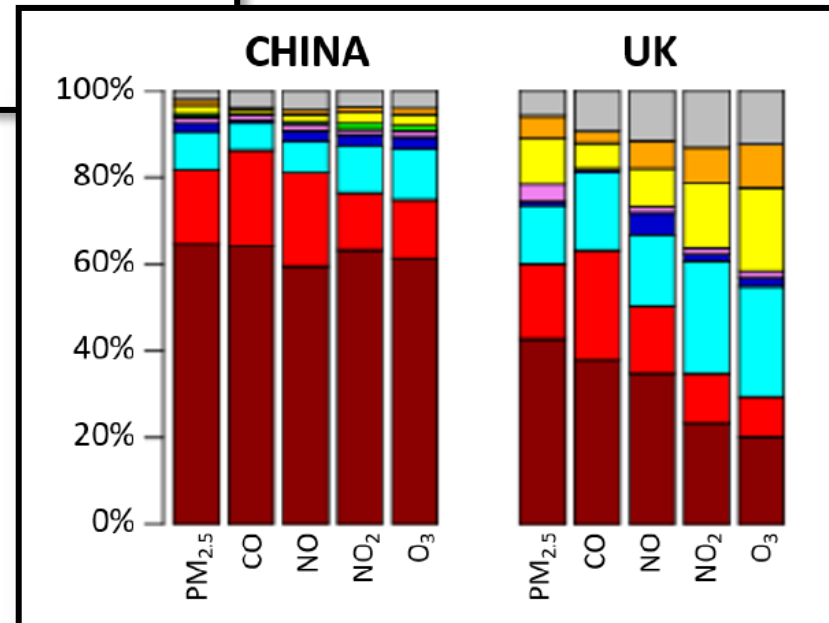
How does this compare to the UK?

(a) Absolute concentrations



Pollutants monitored with PAMs in 250 AIRLESS participants and 35 London participants during a week.

(b) Normalised dose

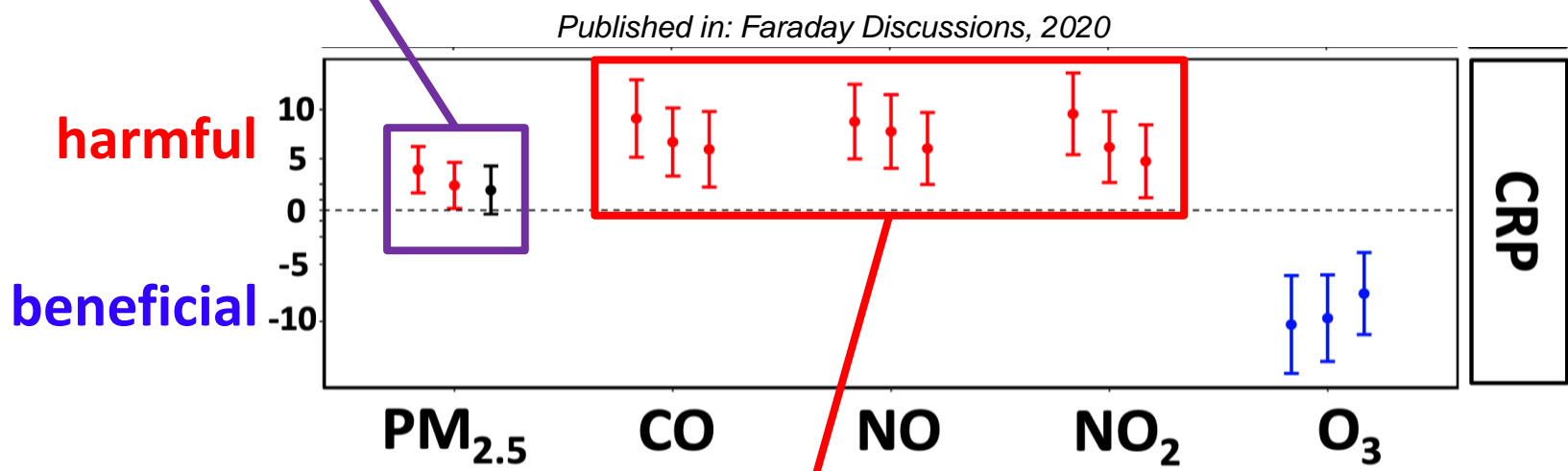


- 1) **Outdoor affects personal exposure**
- 2) **CHINA: home dominates exposure**
- 3) **UK: PM in underground very high**
- 4) **UK: commuting significant fraction**

Relative (health) risks of different pollutants (using ambient air measurements)

Mixed effect linear models
(single pollutant)

Lags of 1,2,3 days



harmful

beneficial

CRP

- **PM_{2.5}, CO, NO and NO₂ all show significant harmful associations (proxies?)**
- **O₃ shows significant beneficial outcome (?)**

C-reactive protein (CRP) is released into the blood within a few hours after tissue injury, the start of an infection or other inflammation.

Analysis and graph by Yiqun Han

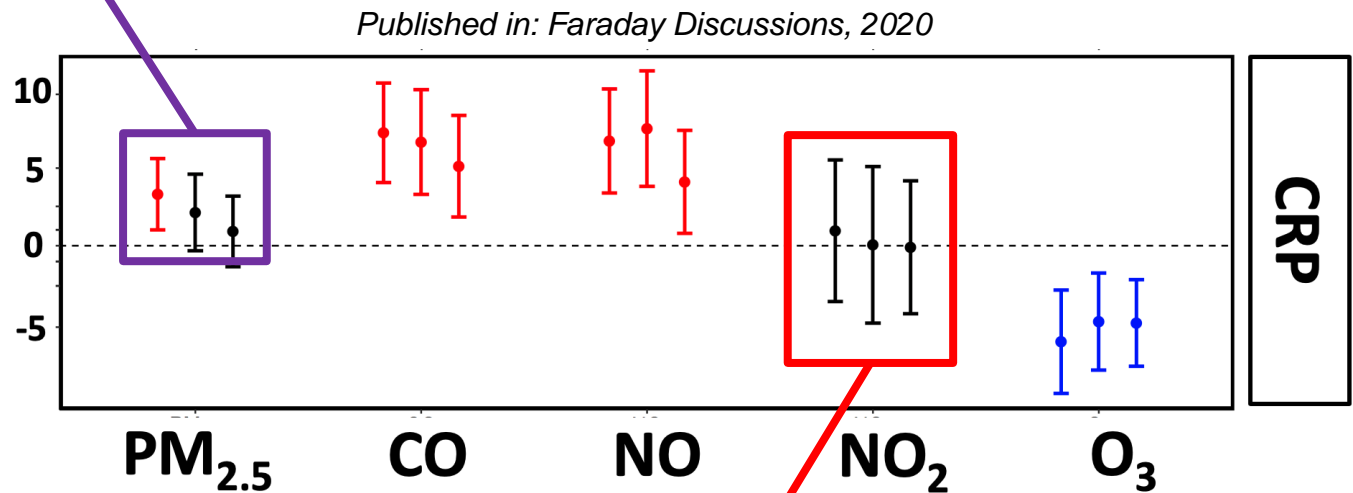
Relative (health) risks of different pollutants (using personal measurements)

Mixed effect linear models (single pollutant)

Lags of 1,2,3 days

harmful

beneficial



CRP

C-reactive protein (CRP) is released into the blood within a few hours after tissue injury, the start of an infection or other inflammation.

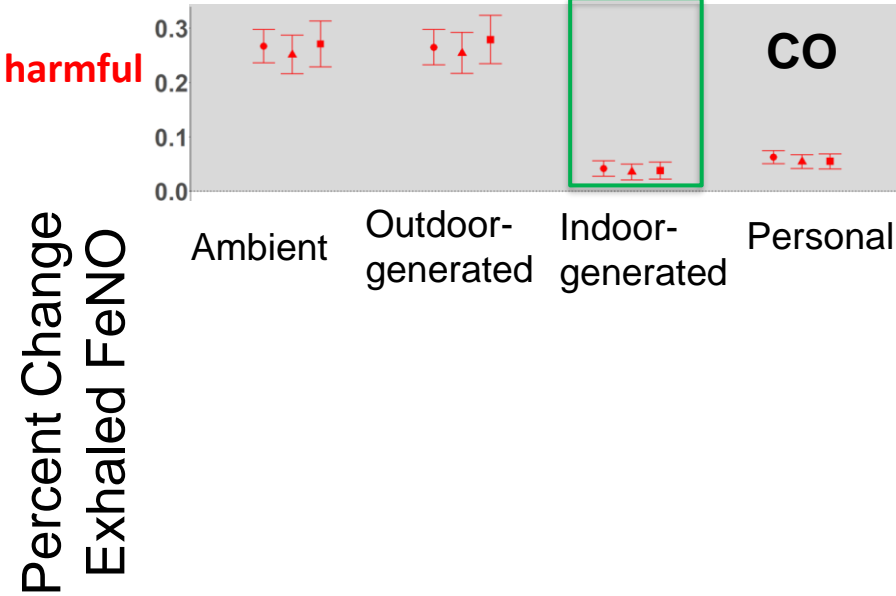
Key difference: NO₂ risk no longer statistically significant.....

Exposure 'error'.....

Analysis and graph by Yiqun Han

Relative (health) risks of different pollutants

Lags of 1,2,3 days



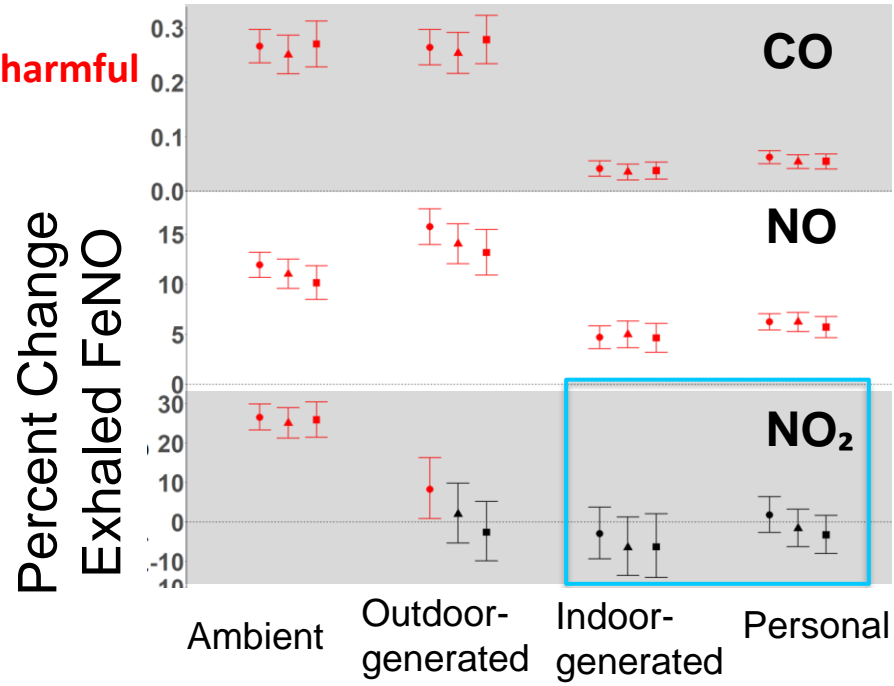
Mixed effect linear models (single pollutant)

⇒ Indoor-generated CO a proxy for indoor combustion sources?

Relative (health) risks of different pollutants

Lags of 1,2,3 days

Mixed effect linear models (single pollutant)

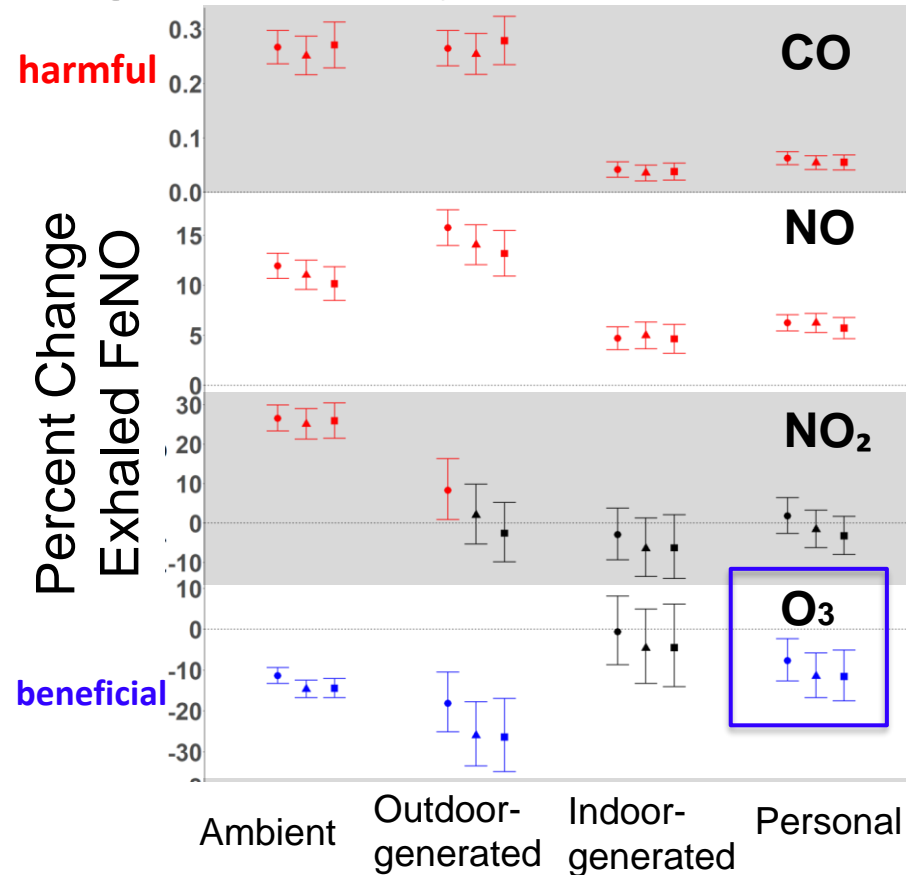


⇒ Indoor-generated CO a proxy for indoor combustion sources?

⇒ Outdoor-generated NO₂ a proxy for traffic intensity

Relative (health) risks of different pollutants

Lags of 1,2,3 days



Mixed effect linear models (single pollutant)

⇒ Indoor-generated CO a proxy for indoor combustion sources?

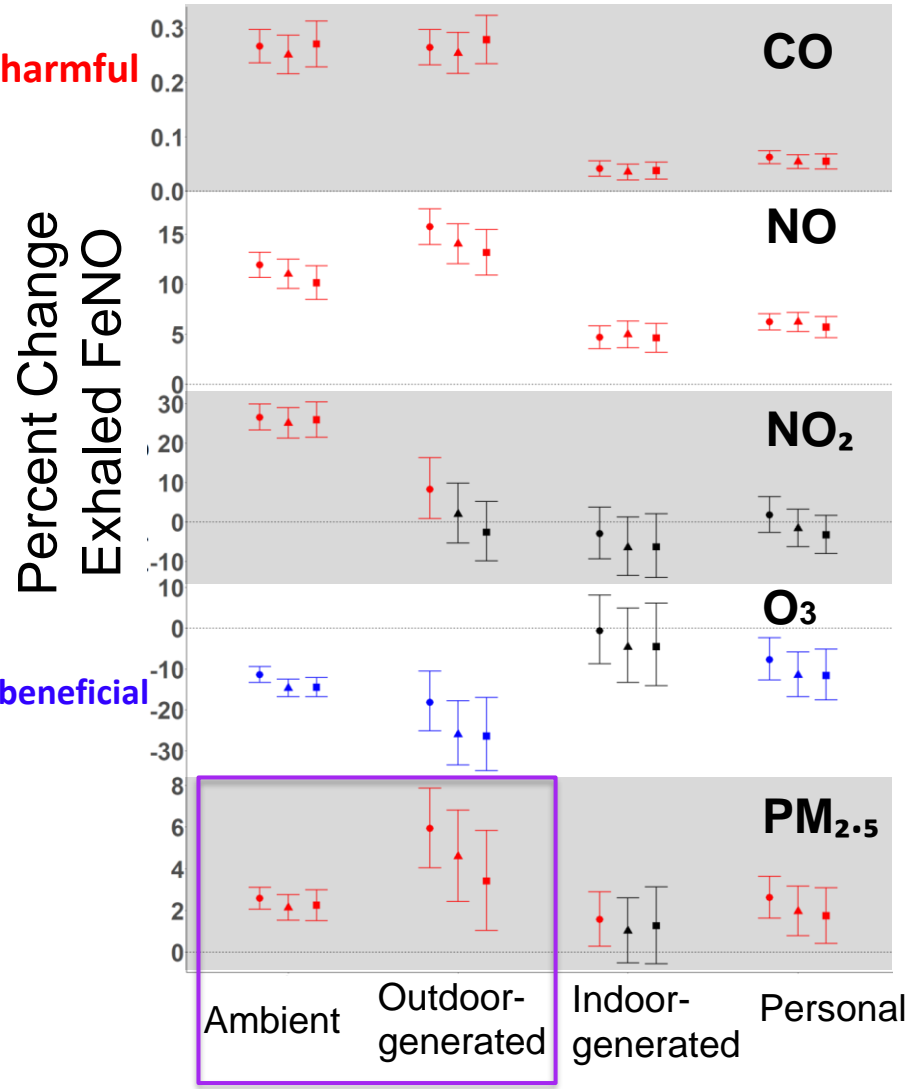
⇒ Outdoor-generated NO₂ a proxy for traffic intensity

⇒ Single-pollutant models cannot control for correlation (or anti-correlation) between pollutants

Relative (health) risks of different pollutants

Lags of 1,2,3 days

Mixed effect linear models (single pollutant)



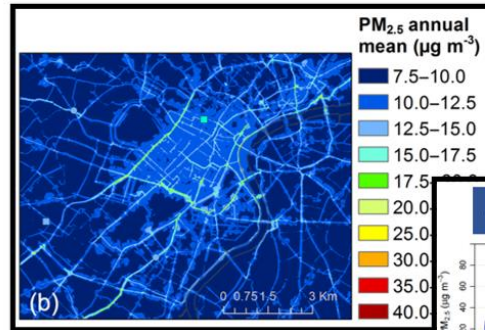
- ⇒ Indoor-generated CO a proxy for indoor combustion sources?
- ⇒ Outdoor-generated NO₂ a proxy for traffic intensity
- ⇒ Single-pollutant models cannot control for correlation (or anti-correlation) between pollutants
- ⇒ Outdoor-generated PM more toxic than previously thought

Personal monitoring vs ambient measurements

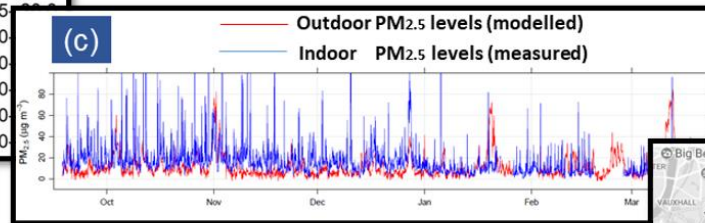
- ⇒ Health risk assessment bias from exposure 'error'
 - ⇒ Effects of source-related exposure on health?
 - ⇒ Improved statistical methods!
- ⇒ understanding and policy implications....

Measurements and models: the next steps

(a) Outdoor dispersion modelling



(b) Indoor air pollution modelling



(c) Activity modelling



⇒ **Extrapolate personal exposure in large-scale health studies to create more reliable health-response functions**

- ⇒ **Advanced scientific knowledge for efficient policy and decision-making**
- ⇒ **Empower individuals/communities to reduce environmental health risks**
- ⇒ **Societal gains**

Dr Lia Chatzidiakou
ec571@cam.ac.uk

+ teams!!

