## Imperial College London Environmental Research Group





Tackling air pollution at school

## Understanding indoor air quality for healthy buildings in a net zero world

## Developing a novel methodological framework for estimating exposure in health and modelling studies (and other stories)

Lia Chatzidiakou ec571@cam.ac.uk Yusuf Hamied Department of Chemistry University of Cambridge



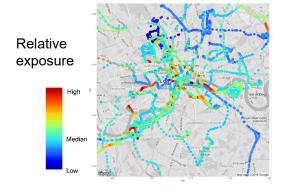
28<sup>th</sup> September 2022 Imperial College London, London Mortality rate attributed to household and ambient air pollution, by WHO region, 2012



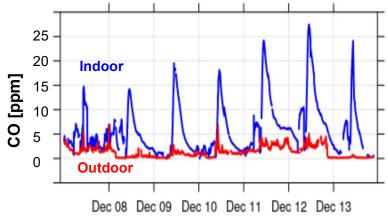
# Air pollution and health

Estimated 8M premature deaths annually globally But....

Activity patterns affect personal exposure and dose Most time spent <u>indoors</u> exposure to indoor air pollution major component

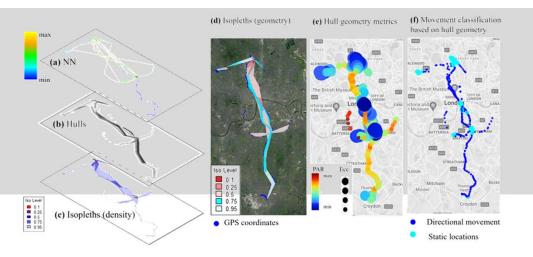


Transit: 6% Other: 10% Home: 84% Indoor air quality (indoor sources, vs outdoor air) very different from ambient pollution





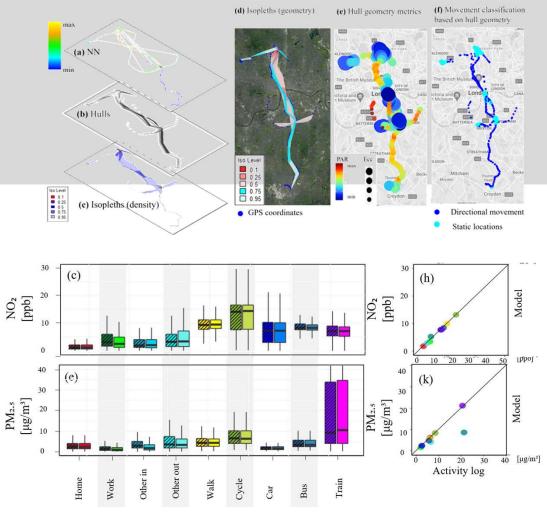
## $\Rightarrow$ Linking activity to dose and to health



## Smart AI/GIS time activity model for exposurerelated activities

### ⇒ Personal activity affects exposure



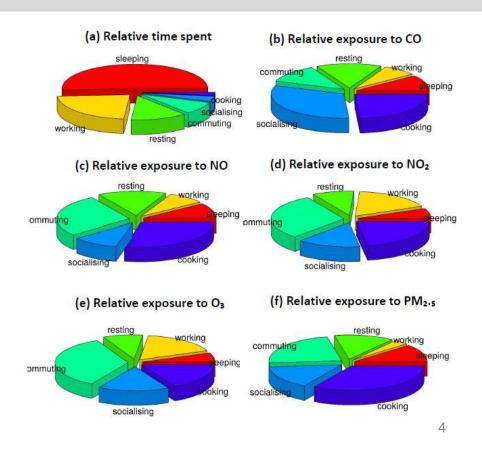


Personal activity affects exposure

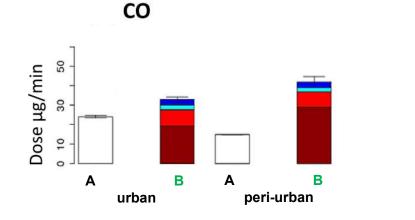
**UNIVERSITY OF** 

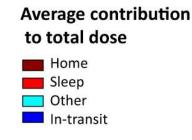
CAMBRIDGE

## Smart AI/GIS time activity model for exposure-related activities: examples



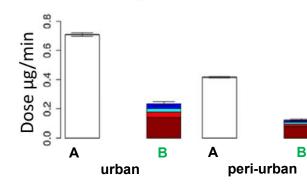
## Personal exposure vs ambient monitors...



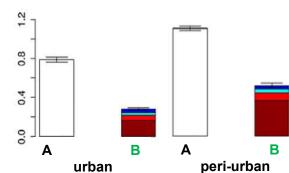












B: personal monitoring + activity



Big differences between personal vs outdoor air quality.....

## Health outcomes (Linear Mixed Models)

### **Ambient**

 $\overline{\text{CO}, \text{NO}, \text{NO}_2}$  all show the same (harmful) effect - why?

 $O_3$  can have a *protective* effect? ...  $O_3$  often <u>anti-</u> <u>correlated</u> with other pollutants

PM<sub>2.5</sub> harmful...

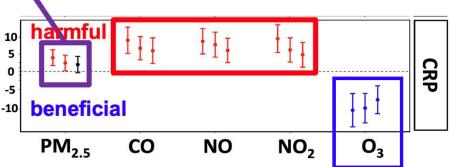
### **Personal**

NO<sub>2</sub> no longer statistically significant.....

Analysis and graph by Yiqun Han UNIVERSITY OF

CAMBRIDGE

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





# Health outcomes (Linear Mixed Models)

### **Ambient**

 $\overline{\text{CO}, \text{NO}, \text{NO}_2}$  all show the same (harmful) effect - why?

 $O_3$  can have a *protective* effect? ...  $O_3$  often <u>anti-</u> <u>correlated</u> with other pollutants

PM<sub>2.5</sub> harmful...

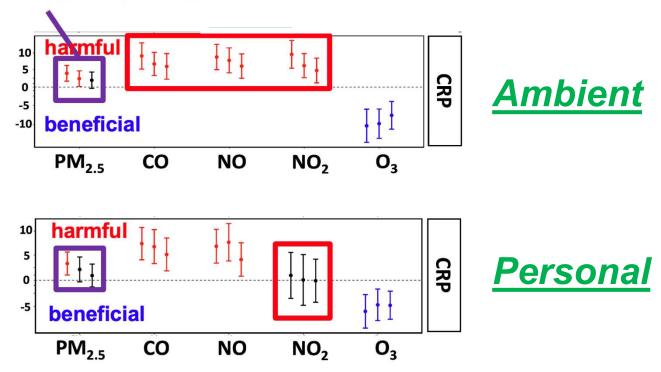
### **Personal**

NO<sub>2</sub> no longer statistically significant.....

Analysis and graph by Yiqun Han UNIVERSITY OF

CAMBRIDGE

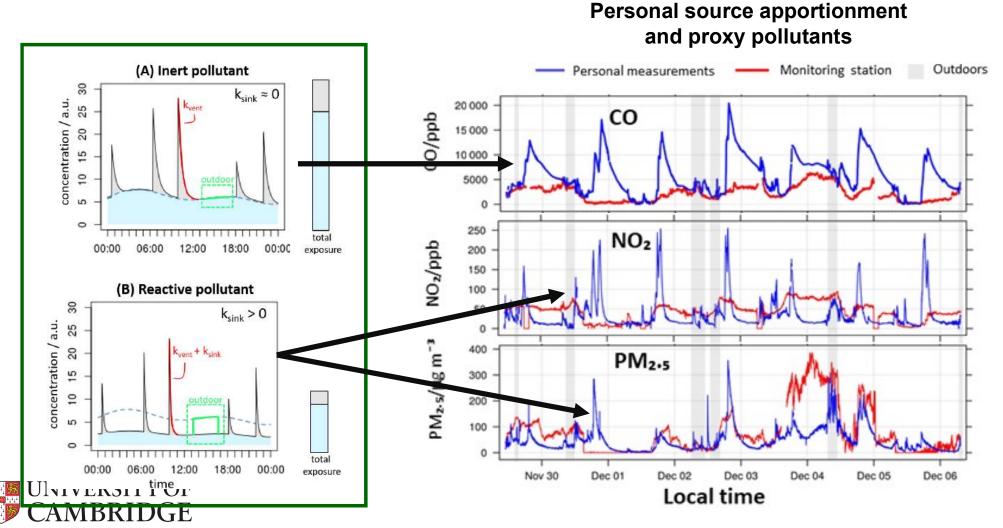
Relative (health) risks of different pollutants Lags of 1,2,3 days



Published in: Faraday Discussions, 2020

*⇒Ambient measurements cannot reliably distinguish causal links* 

# Indoor/outdoor interface – importance of air exchange, chemistry and proxies.....



# School-based exposure (CLASS-ACT)



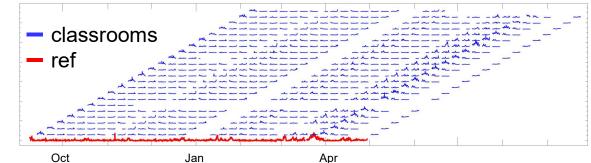
30 schools in Bradford from CLASS-ACT study. Outdoor PM measurements from the closest urban background monitoring station in Leeds (ref) @openair



# School-based exposure (CLASS-ACT)



### 3D plot School C03 PM2.5

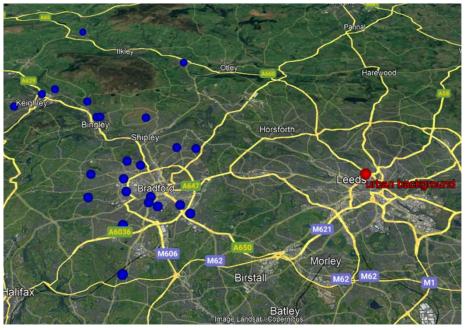


*30 schools in Bradford from CLASS-ACT study. Outdoor PM measurements from the closest urban background monitoring station in Leeds (ref) @openair* 



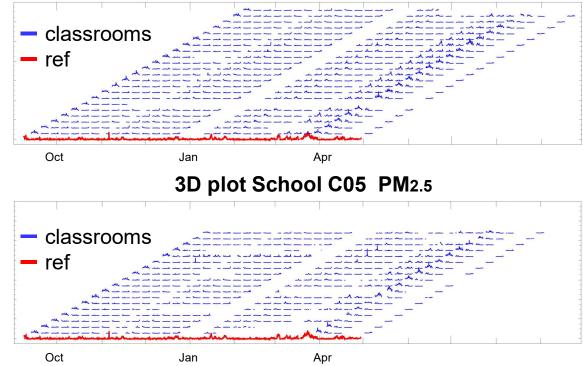
### Classrooms similar PM levels

# School-based exposure (CLASS-ACT)



*30 schools in Bradford from CLASS-ACT study. Outdoor PM measurements from the closest urban background monitoring station in Leeds (ref) @openair* 

### 3D plot School C03 PM2.5





### Schools similar PM levels!!!

# School-based exposure (CLASS-ACT)

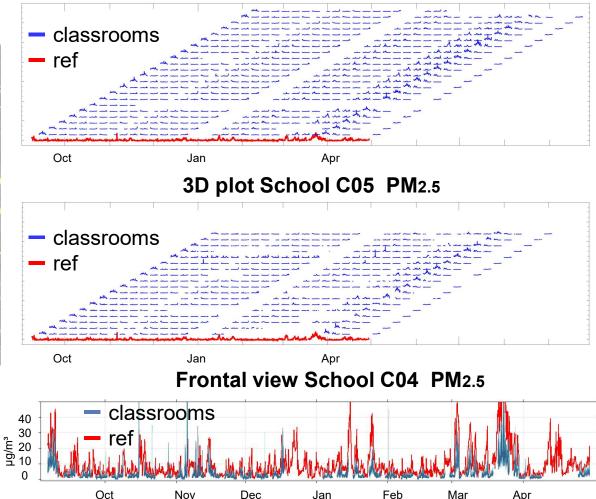


30 schools in Bradford from CLASS-ACT study. Outdoor PM measurements from the closest urban background monitoring station in Leeds (ref) @openair

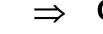
UNIVERSITY OF

CAMBRIDGE

## 3D plot School C03 PM2.5



⇒ Outdoor-generated PM important component



## What does this tell us....

Time-activity patterns important for personal exposure

⇒ Created reliable models for activity classification using ubiquitous data (i.e. smartphones, sensors ....)

## Buildings significant modifiers of personal exposure

⇒ Modelling studies to identify factors affecting indoor exposure and develop novel exposure metrics (i.e. indoor- and outdoorgenerated pollutants, proxies of non-targeted compounds ...)

Exposure error ("misclassification") introduces error and bias in health models

 $\Rightarrow$  Improved exposure metrics and analytical methods

Amazing insights from CLASS-ACT data

 $\Rightarrow$  Evidence of regional-scale PM indoors!



### Selected references

1. Chatzidiakou L. et al., (2019) Characterising low-cost sensors in highly portable platforms to quantify personal exposure in diverse environments, AMT, <u>https://doi.org/10.5194/amt-12-4643-2019</u>

2. Chatzidiakou L. et al., (2022) Automated classification of time-activity-location patterns for improved estimation of personal exposure to air pollution, Environmental Health, *accepted* 

3. Chatzidiakou L. et al., (2020) <u>Using low-cost sensor technologies and advanced computational methods to improve</u> dose estimations in health panel studies: results of the AIRLESS project, JESEE, <u>https://doi.org/10.1038/s41370-020-0259-6</u>

#### **Financial support**

This research has been supported by the NERC (grant no. NE/N007085/1). This research has been jointly supported by the National Natural Science Foundation of China (NSFC grant 81571130100) as well as the Natural Environment Research Council (NERC grant NE/N007018/1) and the Medical Research Council of the UK (AIRLESS project). This research has also been supported by the Medical Research Council of the UK (MR/L019744/1) (COPE project).

CLASS-ACT and SAMHE projects...

