Does the location of sensors matter when monitoring indoor air pollutant exposure in UK homes?

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111

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Health Impact of Air Pollution



Indoor Air Quality

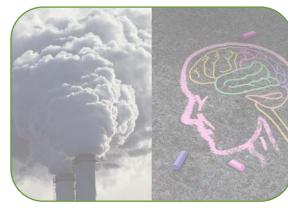


Monitoring



Net Zero

Petrou, G., Hutchinson, E., Mavrogianni, A., Milner, J., Macintyre, H., Phalkey, R., Hsu, S.-C., Symonds, P., Davies, M. and Wilkinson, P. (2022). 'Home energy efficiency under net zero: time to monitor UK indoor air'. *BMJ*, p. e069435. doi: <u>10.1136/bmj-2021-069435</u>.



Health Impact of Air Pollution



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 Analysis

 Home energy efficiency under net zero: time to monitor UK indoor air

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Giorgos Petrou and colleagues argue for systematic large scale monitoring of indoor air to avoid unintended harms to health from home energy efficiency programmes

Net Zero

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Monitoring







Number and location

Petrou, G., Hutchinson, E., Mavrogianni, A., Milner, J., Macintyre, H., Phalkey, R., Hsu, S.-C., Symonds, P., Davies, M. and Wilkinson, P. (2022). 'Home energy efficiency under net zero: time to monitor UK indoor air'. *BMJ*, p. e069435. doi: <u>10.1136/bmj-2021-069435</u>.

Research Question

Does the location of sensors matter when monitoring indoor air pollutant exposure in UK homes?







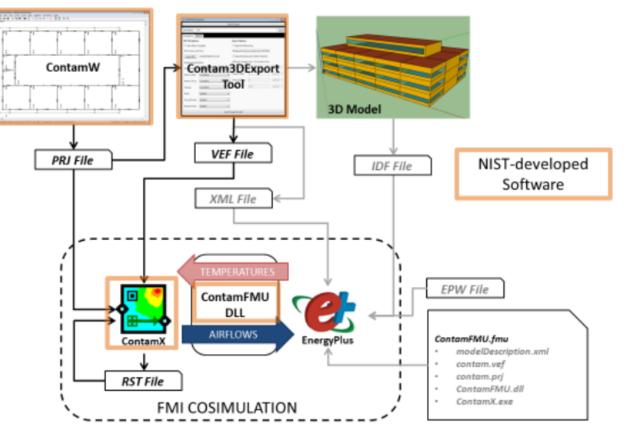


Figure 2. EnergyPlus-CONTAM Coupling Schematic

Dols, W. S., Milando, C. W., Ng, L., Emmerich, S. J. and Teo, J. (2021). 'On the Benefits of Whole-building IAQ, Ventilation, Infiltration, and Energy Analysis Using Co-simulation between CONTAM and EnergyPlus'. in. *8th International Building Physics Conference*, Copenhagen, Denmark, p. 10.



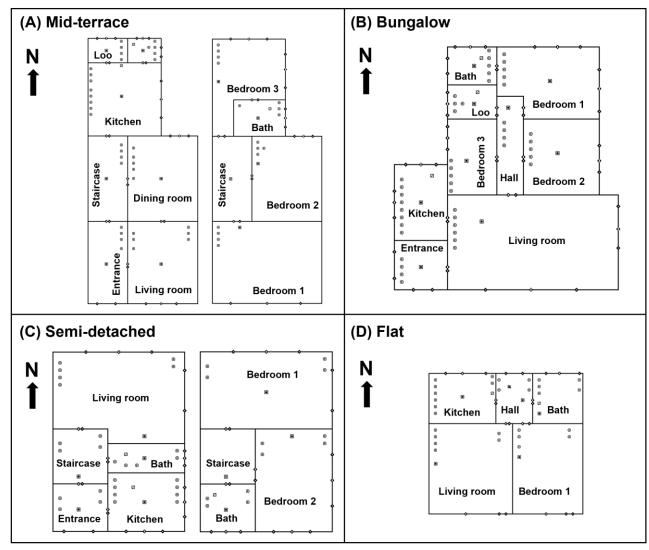
Radon

- Naturally occurring radioactive gas
- Seeps into homes through the floor
- 2nd most important risk factor for lung cancer after smoking

Fine Particulate Matter (PM_{2.5})

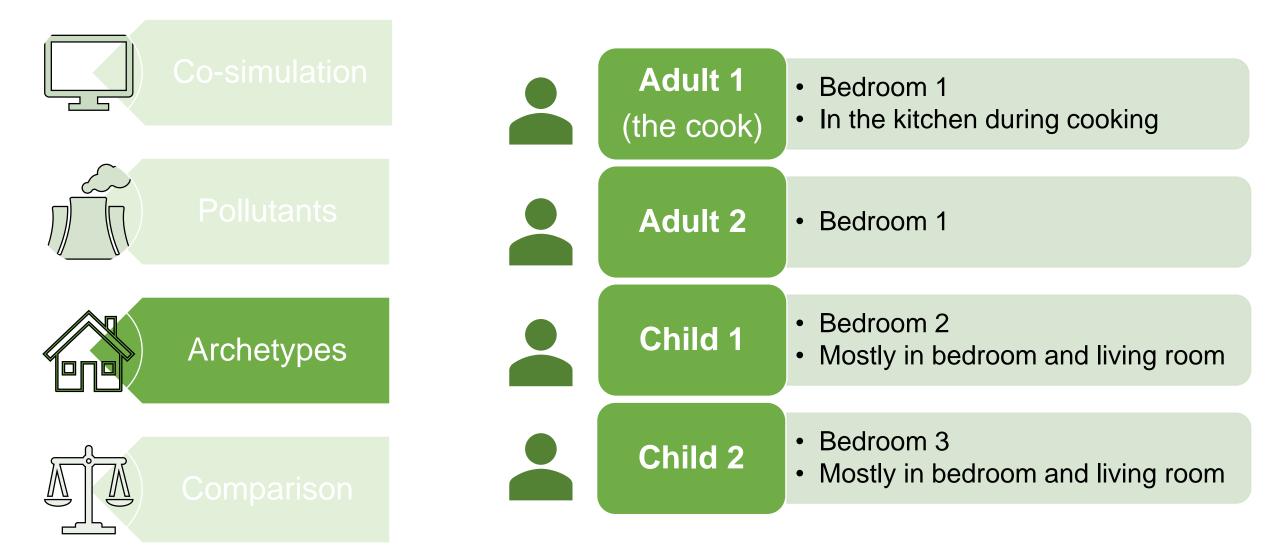
- Indoors sources: cooking, smoking, wood burning stoves.
- Outdoors sources: road transport, industry.
- Coronary heart disease, lung cancer, asthma, [...]

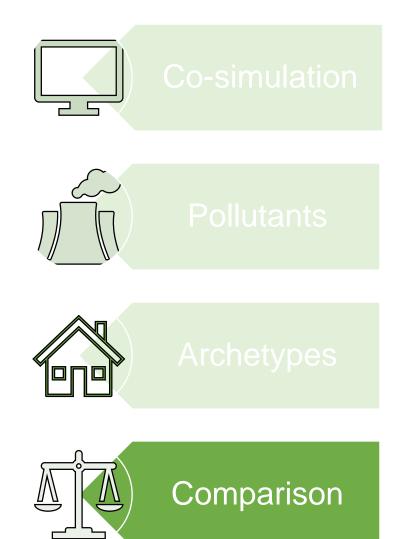




Building layouts based on Oikonomou et al.

Oikonomou, E., Davies, M., Mavrogianni, A., Biddulph, P., Wilkinson, P. and Kolokotroni, M. (2012). 'Modelling the relative importance of the urban heat island and the thermal quality of dwellings for overheating in London'. *Building and Environment*, 57, pp. 223–238.





- The *exact pollutant exposure* for each occupant, as calculated from the full dataset of concentrations in the main rooms (kitchen, living room and bedrooms)
- The *approximate exposure* estimated when data on pollutant concentrations were available for only subsets of the rooms.

Experiment	Living room	Kitchen	Bed. 1	Bed. 2	Bed. 3
К	N	Y	N	N	N
L	Y	Ν	N	N	N
B1	N	Ν	Y	N	N
LB1	Y	Ν	Y	N	N
KB1	Ν	Y	Y	N	Ν

Rooms being monitored per experiment.



$$RMSE(p,i) = \sqrt{\frac{1}{T_i} \sum_{t=1}^{T_i} \left(e_{p,i,t} - e_{p,i,t}^{(a)}\right)^2}$$

$$CV(RMSE(p,i)) = \frac{1}{\overline{e_{p,i}}}RMSE(p,i) \times 100\%$$

$$MBE(p,i) = \frac{1}{T_i} \sum_{t=1}^{T_i} \left(e_{p,i,t} - e_{p,i,t}^{(a)} \right)$$

$$NMBE(p,i) = \frac{1}{\overline{e_{p,i}}}MBE(p,i) \times 100 \%$$

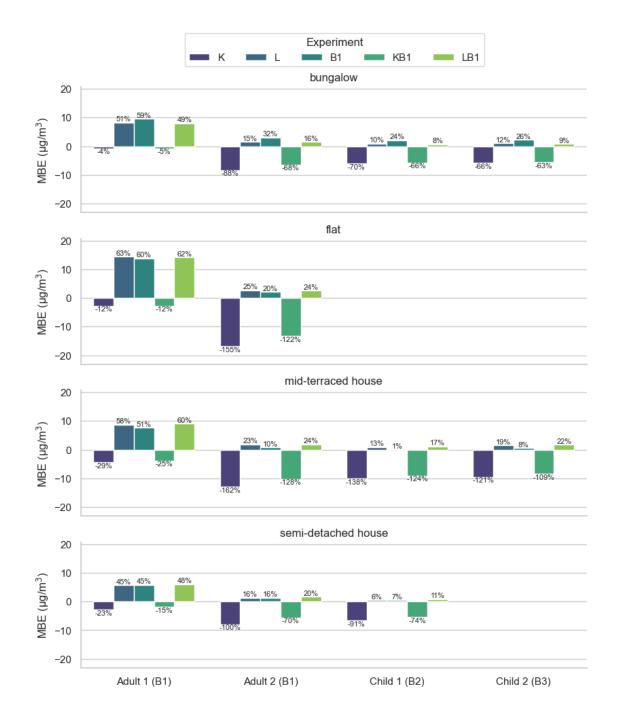
Results



Results – PM_{2.5}

The cook:

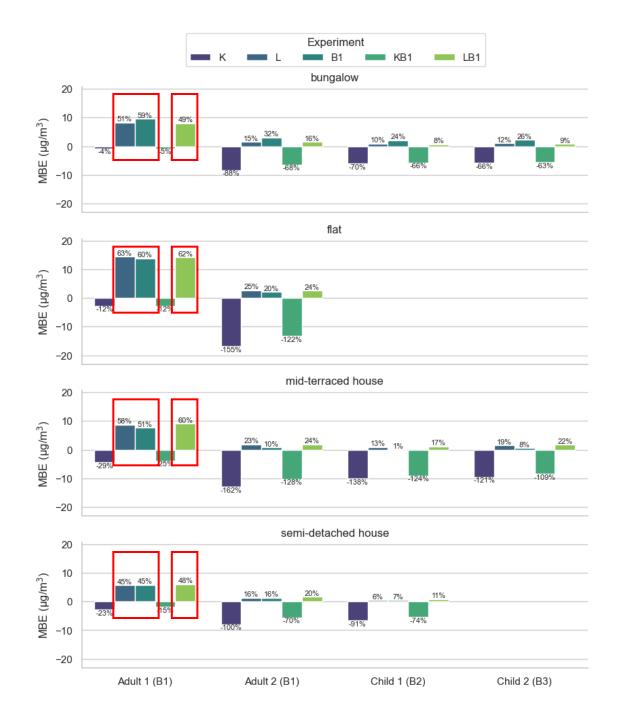
• PM_{2.5} exposure dominated by cooking activities



Results – $PM_{2.5}$

The cook:

- PM_{2.5} exposure dominated by cooking activities
- If data are not collected from the kitchen, we would underestimate their exposure:
 - MBE (NMBE) up to 14.4µg/m³ (63%)



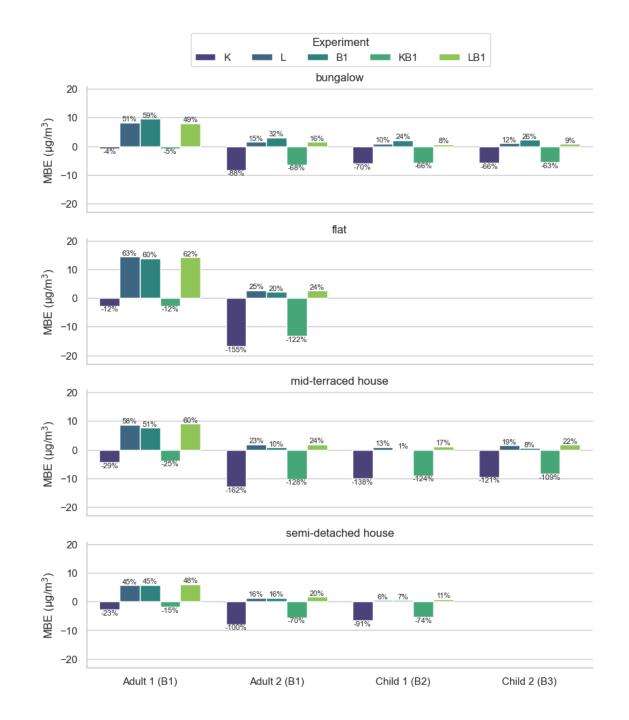
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The cook:

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Other occupants:

Cooking activities have a smaller impact on their overall exposure



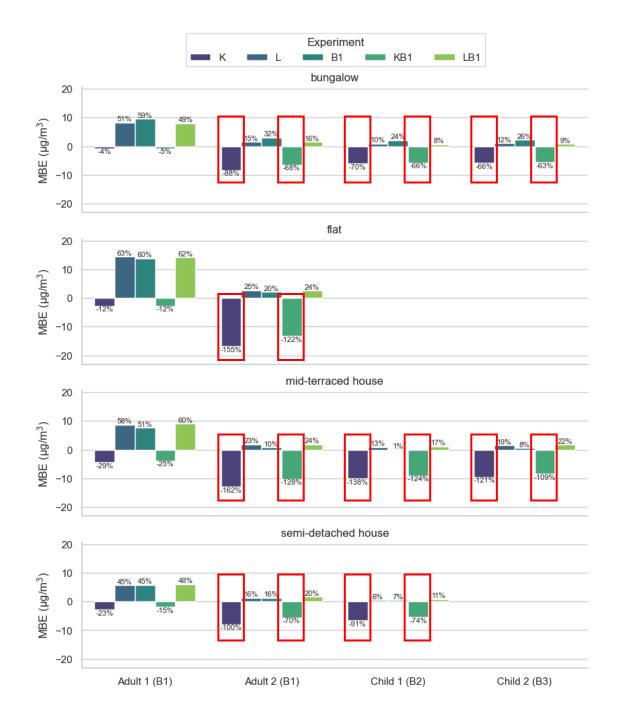
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Other occupants:

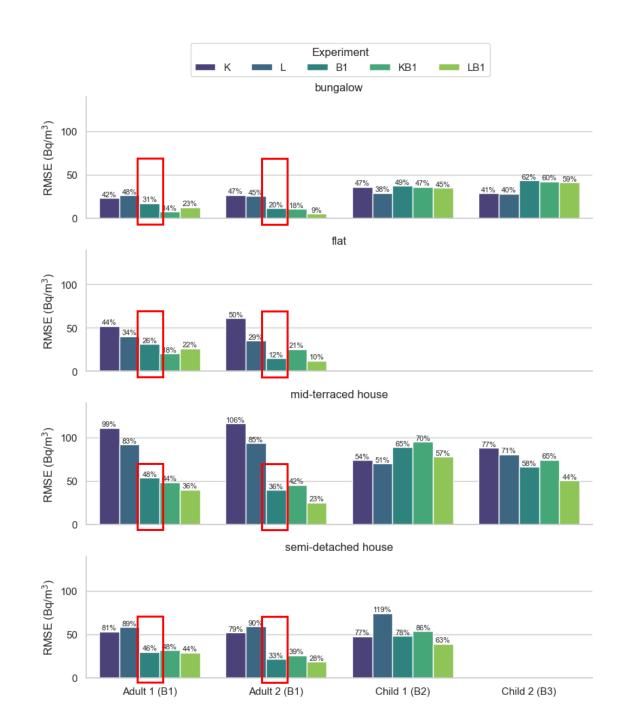
- Cooking activities have a smaller impact on their overall exposure
- Using kitchen measurements only or supplemented with bedroom measurements – would overestimate their exposure:
 - MBE (NMBE) up to 16.8µg/m³ (155%)



Results – Radon

Main bedroom data:

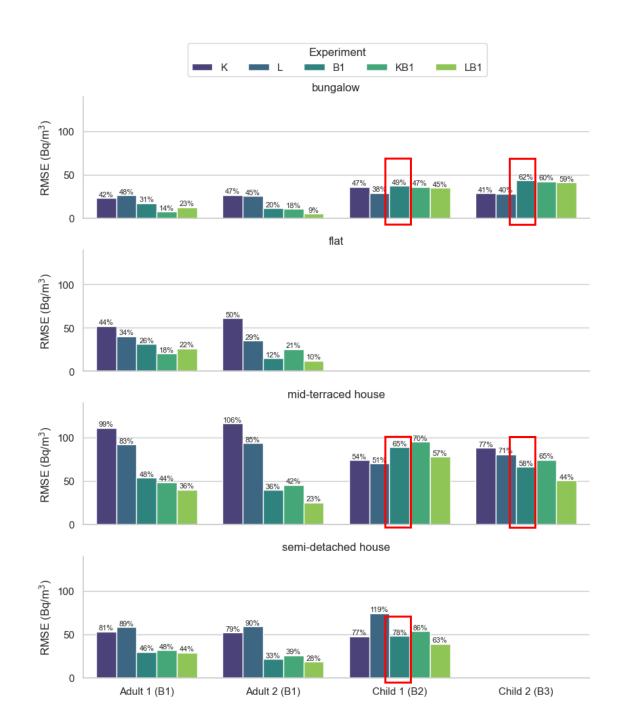
- When using data only from one room, the use of main bedroom data resulted:
 - In the smallest errors for the occupants that sleep in that bedroom



Results – Radon

Main bedroom data:

- When using data only from one room, the use of main bedroom data resulted:
 - In the smallest errors for the occupants that sleep in that bedroom
 - In some cases, in the largest errors for occupants not sleeping in the main bedroom



Discussion



Discussion

Implications

- "Main Bedroom" Generalising based on a subset of rooms can under- or over-estimate exposure
- Comprehensive monitoring
- Where not possible:
 - 1. Identify the typical use per room and by each occupant
 - 2. Identify key pollutant sources
 - 3. Choose sensor placement

Limitations

- Two air pollutants
- Four archetypes
- Limited occupant behaviour diversity
- Well-mixed indoor air

Summary



Summary

- Investigated the impact that sensor placement has on quantifying indoor occupant exposure to PM_{2.5} and radon.
- Key messages:
 - Monitoring a subset of the rooms can result in under- or over-prediction, depending on: sensor location, the occupant, the pollutant, and the archetype.
 - Consider the above factors ahead of detailed monitoring!
- Through further modelling, but also monitoring, **future work** should aim to define best practice for monitoring indoor air quality in homes

Acknowledgments







WK Health Security Agency

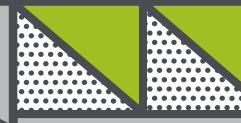
Policy and Implementation for Climate & Health Equity (PAICE)













Thank you for listening!

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Publications & Contact Details



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Methods – Pollutants

Radon:

Used a pressure-driven approach to model ingress of radon through the floor

PM2.5:

- Outdoor-sourced: Hourly data for Plymouth 2019 (Defra Data Archive)
- Indoor-sourced: Cooking
- Other processes: Deposition and windowdependent penetration

Contaminant activity	Rate/Fact or	Schedule
PM _{2.5} emission from cooking	1.6 mg/min	Weekdays: 7.20-7.35, 18-18.30 Weekends: 8.45-9, 12.30-13, 18- 18.30
PM _{2.5} deposition	-0.39/h	24 h
PM _{2.5} penetration factor	0.8	Windows closed
PM _{2.5} penetration factor	1.0	Windows open

Table 1 – $PM_{2.5}$ assumptions based on Shrubsole et al.⁷

[7] Shrubsole, C., Ridley, I., Biddulph, P., Milner, J., Vardoulakis, S., Ucci, M., Wilkinson, P., Chalabi, Z. and Davies, M. (2012). 'Indoor PM2.5 exposure in London's domestic stock: Modelling current and future exposures following energy efficient refurbishment'. *Atmospheric Environment*, 62, pp. 336–343. doi: <u>10.1016/j.atmosenv.2012.08.047</u>.

Methods – Archetype models

- Naturally ventilated with extract fans in the bathroom and kitchen
- Windows open during the summer occupied hours if the indoor temperature exceeded 22 °C
- Bathroom and kitchen windows open during winter when showering or cooking
- Winter heating with a setpoint of 22 °C

Characteristic	Value
Air permeability	15 m ⁻³ h ⁻¹ m ⁻² @50Pa
Wall U-value	1.7 Wm ⁻² K ⁻¹
Window U-value	4.8 Wm ⁻² K ⁻¹
Floor U-value	1.2 Wm ⁻² K ⁻¹
Loft U-value	0.4 Wm ⁻² K ⁻¹
Roof U-value	2.3 Wm ⁻² K ⁻¹

 Table 2 – Building characteristics

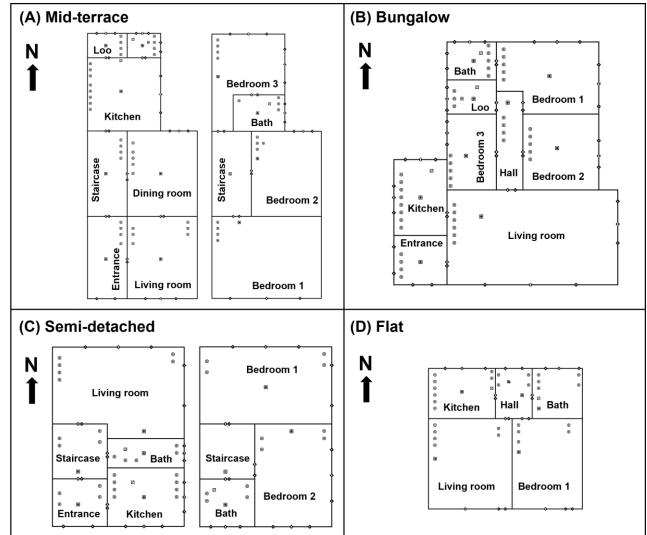


Figure 3 – Building layouts based on Oikonomou et al.⁸

[8] Oikonomou, E., Davies, M., Mavrogianni, A., Biddulph, P., Wilkinson, P. and Kolokotroni, M. (2012). 'Modelling the relative importance of the urban heat island and the thermal quality of dwellings for overheating in London'. *Building and Environment*, 57, pp. 223–238.