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Indoor particulate matter (PM) from cooking in UK students' studio flats and associated intervention strategies: evaluation of cooking methods, PM concentrations and personal exposures using low-cost sensors†

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Cooking emissions have been identified as a major source of indoor particulate matter (PM), which can contribute to severe health issues, including cardiovascular disease and lung cancer. Both cooking methods and use of extractors were investigated to assess their influence on PM concentrations and emission rates. The impact of PM residence times was also examined in terms of the operator's exposure. PM₁₀ and PM_{2.5} were monitored by placing carefully validated low-cost sensors, with sufficiently high accuracies of 70.7% and 64.3% on PM₁₀ and PM_{2.5}, respectively, by compared to a reference instrument, in the kitchen areas of five student studio flats. Ranges and medians of mean concentrations of PM₁₀ and PM_{2.5} of four studied cooking methods ranked (range (median) (µg m⁻³)) deep-frying (62–236 (151); 6–37 (30)), stir-frying (17–176 (63); 2–38 (7)), boiling (6–41 (17); 2–11 (4)) and steaming (7–23 (12); 2–7 (3)), respectively. The impact of using extractors on PM removal rates ranged between 10.5% and 63.0%. Extractors were shown to accelerate the post-cooking pollutant decay, which therefore resulted in shorter residence times. Personal exposure times also varied with cooking method and operator gender. Indoor PM exposure is associated with PM concentration, cooking duration and decay rate. Based on these results, cooking water-based dishes while operating extractors would improve indoor air quality and reduce PM exposure.

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Environmental significance

Domestic cooking plays a critical role in releasing gaseous and particulate pollutants into indoor environments and the outdoor atmosphere. This study tested and deployed low-cost sensors to assess the concentrations, emission rates and decays of particulate matter (PM) from cooking in student studio flats, and examined the results from different cooking methods and the use of extractors. Oil-based methods were found to cause *ca.* six times the emission rates of water-based cooking; extractors significantly increased removal rates and shortened PM residence times, particularly important in residential settings where kitchen and living/sleeping spaces are closely linked; together with successful deployment of validated low-cost sensors, this work enables decision-makers and the general public to improve indoor air quality experiences.

Introduction

Personal exposure to indoor air pollution plays one of the most significant roles in threatening public health around the world.¹ Various pollutants, such as particulate matter (PM), volatile organic components (VOCs) and polycyclic aromatic hydrocarbons (PAHs) from the indoor air have received much attention.

In New Jersey, the concentrations of these contaminants indoor were determined to be higher than those in the outdoor atmosphere.² In many countries people spend 80–90% of their time indoors and the association between airborne particulate matter and adverse effects on human health has been documented by numerous studies.^{3,4} Huang *et al.*⁵ conducted a study on non-smoking Chinese women, with the results demonstrating that the risk of lung cancer had risen due to cooking. This was due to the emissions of particles, especially ultrafine particles which may cause respiratory ailments and damage to the cardiovascular system, reproductive system, and blood system by inhaling over a long time period.⁶ In addition, because of indoor exposure to smoke, Abdullahi *et al.*⁷ suggested that more than one million people died annually from

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chronic obstructive pulmonary disease (COPD). Therefore, indoor air quality has become a focus of public attention.⁸

The emissions of indoor air pollution originate from a wide variety of sources. They differ between buildings and depend on the fuels used for cooking and heating, as well as the personal habits of smoking and use of various consumer products.^{9–11} Emissions from combustion of *e.g.* mosquito-repellent incense, candles and cooking all play critical roles in the formation of indoor air pollution.^{12–14} A study by Abt *et al.*¹⁵ characterising sources of indoor particles in houses in Boston, USA, found that the concentrations of indoor particles were impacted significantly by cooking activities, cleaning, and the movement of people. Cooking, known as a considerable source of air pollution and emissions of odour, can emit micro-solid particles, including reduced sulphur compounds (RSCs), aldehydes, organic acids, VOCs and other gases.^{16,17} Key factors include the choice of cooking oils, duration of cooking, fuel type, food ingredients, cooking temperature, style of cooking and ventilation conditions.^{18,19}

In order to gain a better understanding of the association between cooking and particulate air pollution, the concentrations of particle numbers and size distribution of particles generated during cooking has been assessed in multiple studies. It was found that particles produced by various cooking methods contributed 12–20% to the overall ambient aerosols.²⁰ For instance, an 18 month study conducted by Wallace *et al.*²¹ in a house near Washington DC, USA, found that particle numbers from frying reached 10^{14} after only 15 minutes of cooking with over 90% of the particles in the ultrafine size range. A study conducted by Liu *et al.*¹⁴ suggested that cooking was a significant source of primary organic aerosols (POA) and a potential source to form secondary organic aerosols (SOA); different types of cooking oils were found to release varying amounts of organic aerosols. As a common technique of cooking in numerous cultures, frying with vegetable oils can produce a large quantity of particles and organic gases.^{7,22,23} Dennekamp *et al.*⁶ demonstrated that gas stoves generate high concentrations of particles, while electric rings and grills produce smaller amounts of particles. Noticeably, due to coagulation effects for lower diameter particles, concentration levels of fine particulate ($PM_{2.5}$) were found to drop with the increasing time of heating, as the concentration of coarse particles (PM_{10}) increased.²⁴ While irritations of upper airways were caused by PM_{10} and total suspended particulates (TSP), fine particulate matter were osmosing into bronchi where they might ultimately enter the bloodstream.²⁵ The most critical parameter impacting evaporative emission rates was found to be temperature of cooking as it increased particle emission rates.¹⁹

A positive correlation between the emission of VOCs, which is key for formation of particulate matter from heated cooking oils, and frying temperature has been found.²⁶ During oil heating, $PM_{2.5}$ was released with a linkage to emission of heavy metals and organic compounds.^{27,28} Additionally, in terms of the raw material, food containing a high percentage of fat was likely to produce a much higher amount of $PM_{2.5}$ compared to vegetables which contain less fat, with a ratio of peak mass concentration to background concentration for fatty food and

vegetables at approximately 40 and 8, respectively.²⁹ Lu *et al.*⁸ conducted a study of $PM_{2.5}$ emissions by six types of cooking styles using gas as the fuel. Their results showed that emission rates of $PM_{2.5}$ decreased in the following order: deep-frying, stir-frying, stewing, quick-frying, boiling and steaming. Also, See and Balasubramanian³⁰ assessed the exposure to indoor aerosols with the association with Chinese cooking methods on gas-phase by controlled experiments in a domestic kitchen. Their results found that the highest particles concentration was produced by deep-frying, with 20 nm of the mode diameter approximately. However, Zhao *et al.*¹⁸ observed that in Chinese cooking, stir-frying and pan-frying caused a significantly higher generation rate of $PM_{2.5}$ than deep-frying, although it required larger amounts of cooking oil and longer durations of cooking. During the cooking process at a high temperature, it can generate more than 300 reaction products, including fatty acids, alkanes, alkene, aldehydes, ketones, alcohols, esters, aromatic compounds, and heterocyclic compounds.³¹ In order to manage indoor air pollution, in addition to controlling the source of emission, the removal of the pollutants also plays a significant role.³² The use of an extractor directly influences the level of $PM_{2.5}$ and an extractor operating at higher air flow rates has a more significant effect on reducing $PM_{2.5}$ concentrations during cooking.

The majority of previous studies on cooking emissions have been conducted in places such as restaurants, canteens, laboratories and residential kitchens.^{33,34} Also, a large section of the existing literature focusses on emission rates to the ambient air which differs from the indoor environment investigated in the present study.¹⁹ To our knowledge, no previous studies have focused on the cooking emissions within the confined area of student studio accommodation, where young adults are exposed to cooking emissions from its generation through to its decay. In terms of the cooking methods, there are a wide variety of cuisines around the world, associated with geographical environments, cultural traditions, different climates, ethnic customs and other factors. With the increasing number of international students studying at UK universities, students living in university accommodation generally cook for themselves with cuisines and cooking styles showcasing the features from their hometowns. According to a survey conducted by Lu *et al.*,⁸ deep-frying, frying, stir-frying, boiling, stewing and steaming are the most favoured cooking styles.

In the present study, measurements of cooking emissions were carried out in five student studio flats. In such studio flats, students from all over the world dwell together with no partitioning between their private living areas and the kitchen; this common set-up may cause a particularly long exposure to the air pollutants produced by cooking activities. A low-cost sensor was used to record these cooking emissions. The concentrations and emission rates of PM, including PM_{10} and $PM_{2.5}$, from different cooking methods was assessed as well as the intervention by using the kitchen extractor. The link between the cooking emission concentrations and cooking duration was also examined. The efficiency of the kitchen extractor and residence time of the pollutants was also assessed. Based on the results from the measurements, the personal exposure to the



indoor particles from the cooking emissions was determined to estimate potential health impacts.

Experimental

Materials and methods

Study sites. Five kitchens within student studio flats near the University of Birmingham, Birmingham, UK, were chosen for the trials to ensure the sampling sites were under similar meteorological and outdoor air quality conditions, which will influence the indoor concentration of pollutants through air ventilation and penetration processes; the volumes of the flats were of 35.84 m³, 42.56 m³, 42.84 m³, 46.76 m³ and 50.96 m³, and their locations are illustrated in Fig. S1.† The sampling points were at 1.6 m height, *i.e.* where people breathe, since the respiratory area was the location of interest. All the hearths were ceramic electric hobs at a height of 900 mm, and they were well-cleaned before the study. In each kitchen, there was the same extractor fan (model Cookology CH600SS extractor fan with a width of 600 mm, depth of 475 mm and height of 800 mm; it had three air flow settings: mode 1: 150 m³ h⁻¹; mode 2: 300 m³ h⁻¹; mode 3: 450 m³ h⁻¹) located at a distance of 650 mm above the hob.

Sampling instrument. The Flow2, a low-cost air quality sensor (*ca.* £140 in 2021; see Fig. S2†) designed and produced by Plume Labs, was used for data collection. It tracks and records air pollutants including nitrogen dioxide (NO₂), VOCs and PM in its immediate surroundings in one-minute intervals (we only used PM data for the present study). PM is blown into the sensor by an in-built fan and monitored using a laser beam. Every time a particle is hit by the laser, the light is dispersed, which is then detected by a photovoltaic cell that transforms the deflected laser beam into an electrical current that is measured. An automatic calibration happens every time when the sensor and the app on personal devices synchronise over Bluetooth in the background by artificial intelligence (AI). The Flow2 devices were tested at the Plume Labs' Paris headquarters in 2019, where an AeroTrak 9306 Handheld Particle Counter was used as the reference to assess the accuracy of the PM measurement reporting highly correlated results with average correlations for PM₁, PM_{2.5} and PM₁₀ of 92.8%, 92.2% and 88.2%, respectively.³⁵ Research conducted in 2022 examined 34 Flow sensors in comparison to a Plantower A003 reference at John Hopkins University with *R*² values reported for PM_{2.5} and PM₁₀ of 76% and 73%, respectively.³⁶

To confirm the accuracy of the Flow2 measurements for our work, two Flow2 sensors were placed in an office in the Biosciences building at the University of Birmingham with co-location of the reference instrument Fidas® 200 (Palas GmbH), a continuous ambient air quality monitoring system, to assess the inter-Flow2 variation. In line with similar research,^{37,38} the coefficient of variation (CV), which determines the variation among the sensors, the relative precision errors (RPE), which is averaged to indicate the precision of the sensors, and the accuracy (Acc) by comparisons between Flow2 and Fidas® 200 were calculated and linear regressions (using SPSS; IBM version 28) were chosen to yield the *R*² correlation between

the two Flow2 sensors and between each Flow2 sensor and the Fidas® 200 reference instrument (see ESI, Text S3, eqn (S1)–(S3) for details†). The results of averaged CV of PM_{2.5} and PM₁₀ concentrations between the two Flow2 sensors were 12.3% and 15.0%, respectively, and the overall RPE were 17.4% and 21.2%, which demonstrated high precisions between Flow2 sensors. In addition, the overall accuracies of Flow2 sensors by compared to the Fidas® 200 for PM_{2.5} and PM₁₀ were 70.7% and 64.3%, respectively. The averaged values of *R*² for linear regressions between Flow2 sensors and Fidas® 200 for PM_{2.5} and PM₁₀ were 89.5% and 73.4%, respectively (compare Table 1). The moderately high values of the accuracy and coefficient of determinations illustrate suitable performance in terms of reliability for using the Flow2 sensors for further utilisation on indoor air quality assessment focussing on PM_{2.5} and PM₁₀ (it should be noted that we have not tested any of the other pollutants Flow2 attempts to measure).

Sampling strategy. For measuring PM concentrations, the sampling points were set next to the hearth with a distance of 300 ± 50 mm horizontally from the hobs and above the kitchen countertops at a height of 1600 ± 50 mm (see in Fig. 1) which is the nose height for the average stature of humans around the world.³⁹ The horizontal distance between the hob and the Flow2 was chosen to avoid immediate emissions of oil drops and the vertically ascending fumes extracted by the cooker hoods, which may cause damage to the sensor or block the air inlets. The sensors were placed at least 100 mm away from the wall and other surfaces to allow air to circulate freely. In each study kitchen, one calibrated and tested sensor was used. The research for studying the correlations between the concentration of PM and the cooking durations and cooking methods, and the emission rates of different cooking methods lasted ten days in the five kitchens (10th to 20th May 2021). The typical examples of cooking processes to study the relationships between PM concentrations and the usage of kitchen extractors were collected during June 2021 in the kitchen of one of the studied student studio flats.

The Flow2 was set up at least 10 days before the data collection period and synchronised to the apps on personal devices at least once a day for auto-calibration by the AI. During the period for measurements of the PM concentrations, the use of aerosol sprays, laser printers, cleaning and combustion processes such as burning candles and smoking cigarettes, as well as the use of e-cigarettes were prohibited in the kitchens, as these activities may impact on the PM concentration.⁴¹ Before cooking, the hobs and the bottom of the pots were well-cleaned to minimise the influences of the residua on the hearth and pot, which could potentially release additional PM and VOCs while

Table 1 Correlation coefficients (*R*²) for the results of PM concentrations from the two Flow2 sensors (Flow2_A and Flow2_B) and the Fidas® 200 (Fidas), the reference instrument

	Flow2 _A vs. Flow2 _B	Flow2 _A vs. Fidas	Flow2 _B vs. Fidas
PM ₁₀	0.936 (<i>p</i> < 0.001)	0.771 (<i>p</i> < 0.001)	0.697 (<i>p</i> < 0.001)
PM _{2.5}	0.836 (<i>p</i> < 0.001)	0.873 (<i>p</i> < 0.001)	0.908 (<i>p</i> < 0.001)



$$C_t = C_p e^{-\lambda t} \quad (4)$$

where C_p is the PM concentration ($\mu\text{g m}^{-3}$) at peak time, and C_t is the PM concentration ($\mu\text{g m}^{-3}$) at time t (min) after the measured peak concentration.^{43,44} The emission rate S_p was obtained by nonlinear fitting of the increase of indoor particle concentrations based on the real-time concentration $C_{\text{in,p}}(t_0)$, result of air change rate a , the total removal rate λ , and the volume of the kitchens, V .^{40,42}

Removal performance of the kitchen extractors. The kitchen extractors were run during the period of cooking for most cases in real-life scenarios. In order to determine the removal effect of kitchen extractors on the air pollutants generated by cooking, the emission rates of air pollutants with the extractor hoods on would be assessed as well. Again, it was assumed that the air was well-mixed with steady ambient concentration of air pollutants. The equation of mass balance for PM with the kitchen extractors on can be expressed as:

$$\frac{dC_{\text{in,p}}(t)}{dt} = \frac{Q_{\text{ex}}}{V} PC_{\text{out}} - (\lambda - a)C_{\text{in,p}}(t) + \frac{S_p - Q_{\text{ex}}C_{\text{in,p}}(t)}{V} \quad (5)$$

where Q_{ex} is the exhaust air volume rate ($7.5 \text{ m}^3 \text{ min}^{-1}$) of the kitchen extractors, $S_{\text{hood,p}}$ is the emission rate ($\mu\text{g min}^{-1}$) of particulate matter when the kitchen extractors are on.⁴⁰ Then, $S_{\text{hood,p}}$, the surveyed emission rate of particulate matter with the kitchen extractors on is:

$$S_{\text{hood,p}} = S_p - Q_{\text{ex}}C_{\text{in,p}}(t) \quad (6)$$

The calculated emission rate of PM with kitchen extractors on, $S_{\text{hood,p}}$, is affected by the PM concentration in the exhausted air, which varies over time. To make comparisons between this and the emission rates of the pollutants generated from cooking, S_p , the emission rate of PM with the extractors on, $S_{\text{hood,p}}$, is estimated as a constant from the solution of the concentration of indoor PM for eqn (1):

$$C_{\text{in,p}}(t) = \left[C_{\text{in,p}}(t_0) - \frac{\lambda Q_{\text{ex}}}{a(\lambda - a)V} C_{\text{balance}} - \frac{S_{\text{hood,p}}}{(\lambda - a)V} \right] e^{-(\lambda - a)\Delta t} + \frac{\lambda Q_{\text{ex}}}{a(\lambda - a)V} C_{\text{balance}} + \frac{S_{\text{hood,p}}}{(\lambda - a)V} \quad (7)$$

and the calculation of emission rates $\overline{S_{\text{hood,p}}}$ of PM with kitchen extractors on could be rewritten as:

$$\overline{S_{\text{hood,p}}} = \frac{(C_{\text{in,p}}(t) - C_{\text{in,p}}(t_0) e^{-(\lambda - a)\Delta t})(\lambda - a)V}{1 - e^{-(\lambda - a)\Delta t}} - \frac{\lambda Q_{\text{ex}}}{a} C_{\text{balance}} \quad (8)$$

Therefore, the removal efficiency of the kitchen extractors can be expressed as:

$$\eta = \left(1 - \frac{S_{\text{hood}}}{S} \right) \times 100\% \quad (9)$$

where S_{hood} is the emission rate surveyed with the kitchen extractors on, and S is the emission rate of the cooking-generated air pollutants with the extractors off. The windows

and doors were closed or open only with a small acute angle to minimise the influence of outdoor ambient air and ensure the air was well mixed indoors.⁴⁰

Personal exposure assessment for estimation of health effects. Particulate matter plays a critical role in endangering the respiratory and cardiovascular systems directly. The impacts of cooking activities on human health could be reflected and explained by personal exposure assessments of particulate matter in the respiratory zone for people in the student accommodation. Based on the results from the surveys recording cooking activities from the volunteers, eqn (10)–(12) were used to obtain the annual PM inhalation exposures as follows:^{8,45}

$$C_{\text{average}} = \sum C_i / N \quad (10)$$

where C_{average} represented the average concentration ($\mu\text{g m}^{-3}$) of PM from the surveyed cooking processes, C_i was the average concentration ($\mu\text{g m}^{-3}$) of PM per dish, and N was the number of dishes.

$$t_e = \sum_{i=1}^N (t_1 - t_0) / N \times \bar{N} \times f_c \quad (11)$$

where t_e represented the exposure time (min) per day, t_1 and t_0 represented the timepoints of starting and ending of each cooking activity, \bar{N} represented the average number of dishes per cooking activity, f_c represented the averaged frequency of cooking activities per day.

$$D_{\text{pot}} = C_{\text{average}} \times \text{IR} \times t_e \times \text{EF} \quad (12)$$

where D_{pot} was the annual PM intake (μg per year), IR was shorted for inhalation rate, and EF was the exposure frequency.⁸ According to Wang and Duan,⁴⁶ the average inhalation rates of male and female people aged between 18–40 years old with moderate activities were $0.66 \text{ m}^3 \text{ h}^{-1}$ and $0.59 \text{ m}^3 \text{ h}^{-1}$, respectively, due to the physical differences. The exposure frequency was assumed to be 300 day per year.

Results and discussion

PM emissions from different cooking methods

The ranges and medians of mean concentrations of PM_{10} and $\text{PM}_{2.5}$ generated by the four cooking methods deep-frying, stir-frying, boiling and steaming under conditions of both kitchen extractors on and off in the five study kitchens are displayed in order from high to low emission in Tables 2 and 3. The minimum concentrations for PM_{10} and $\text{PM}_{2.5}$, *i.e.*, the background pollution levels were $3.0 \mu\text{g m}^{-3}$ and $2.0 \mu\text{g m}^{-3}$, respectively. The maximum concentrations of PM_{10} and $\text{PM}_{2.5}$ peaked at $2111.5 \mu\text{g m}^{-3}$ and $57.4 \mu\text{g m}^{-3}$, respectively, during deep-frying. Statistically significant differences in the mean concentrations of PM_{10} and $\text{PM}_{2.5}$ between different cooking methods were found ($p < 0.001$). The average concentrations of PM_{10} and $\text{PM}_{2.5}$ in the direct respiratory zone during the cooking process were 48.7 and $9.4 \mu\text{g m}^{-3}$, respectively. This indicated that the average levels of the cooking emissions were



Table 2 Ranges and median of PM₁₀ and PM_{2.5} mean concentrations (C, $\mu\text{g m}^{-3}$) and duration (min) of cooking activities of each cooking method (range (median))

	$C_{\text{PM}_{10}}$	$C_{\text{PM}_{2.5}}$	Duration
Deep-frying	62–236 (151)	6–37 (30)	7–49 (16)
Stir-frying	17–176 (63)	2–38 (7)	10–61 (23)
Boiling	6–41 (17)	2–11 (4)	9–172 (17.5)
Steaming	7–23 (12)	2–7 (3)	7–65 (21)
Total	6–236 (30)	2–38 (5)	— ^a

^a The column of duration indicated the range of time taken for each dish by each cooking method, so the row of total would not determine any useful value. According to the survey, the average cooking duration per day was 49.4 minutes.

Table 3 Ranges and medians of mean concentrations of PM₁₀ and PM_{2.5} of cooking activities of each cooking method with or without the extractors on (range (median)) ($\mu\text{g m}^{-3}$)

	Extractors off		Extractors on	
	$C_{\text{PM}_{10}}$	$C_{\text{PM}_{2.5}}$	$C_{\text{PM}_{10}}$	$C_{\text{PM}_{2.5}}$
Deep-frying	91–236 (190)	22–35 (27)	62–216 (146)	10–37 (34)
Stir-frying	52–176 (67)	5–38 (12)	17–129 (54)	2–22 (6)
Boiling	9–41 (17)	2–9 (4)	6–35 (12)	2–11 (4)
Steaming	7–23 (12)	2–7 (4)	7–14 (11)	2–4 (3)

around or below the WHO air quality guideline values for PM₁₀ and PM_{2.5} for the 24 hour mean (45 and 15 $\mu\text{g m}^{-3}$, respectively).⁴⁷ However, the average values of PM included boiling and steaming, which emitted particles at very low levels and lasted longer. The concentrations of PM₁₀ emitted from deep-frying and stir-frying exceeded the PM₁₀ level recommended by the UK Department for Environment Food & Rural Affairs,⁴⁸ with the standards of annual mean PM₁₀ and PM_{2.5} at 40 $\mu\text{g m}^{-3}$ and 25 $\mu\text{g m}^{-3}$, respectively. Approximately 73% of cooking activities in the studied flats recorded peaks in PM₁₀ concentrations higher than the standard, and 67% of them lasted for more than 10 minutes. Even though the average PM_{2.5} concentrations did not go beyond the standard, 45% of the surveyed cooking activities recorded peaks above 25 $\mu\text{g m}^{-3}$, with 28% lasting for more than 10 minutes. Noticeably, the mass concentrations of both PM₁₀ and PM_{2.5} from deep-frying and stir-frying were greater than those from boiling and steaming. The fact that cooking periods and dishes cooked varied introduced significant uncertainties. Thus, ranges and medians were used to present the concentrations and emission rates of the cooking activities.

Several studies have demonstrated that compared to low-fat foods, high-fat foods generate more particulate matter during the process of cooking, and oil-based cooking methods, including deep-frying, stir-frying, pan-frying and grilling emit more particles, especially fine particles, than cooking methods that are water-based, including boiling, stewing and steaming.^{30,49–51} This is likely due to the enhanced oil and food pyrolysis caused by the high temperature, which results in much higher particle emissions from pan-, deep- and stir-frying

than those of boiling and steaming.¹⁹ According to To and Yeung,⁵² by combining the deep-, stir- and pan-frying and the other methods of cooking, the number concentration of PM₁₀ rose to up to $3.9 \times 10^7 \text{ cm}^{-3}$, while PM_{2.5} concentrations were discovered to be elevated 90 times higher than the background levels during the cooking process of grilling and frying.⁵² The cooking process of stir-frying was observed to generate the highest concentration of PM₁₀ with a peak at 1300 $\mu\text{g m}^{-3}$, followed by pan- and deep-frying in domestic kitchens.⁵² See and Balasubramanian³⁰ conducted experiments by cooking 150 g of tofu *via* steaming, boiling, stir-frying, pan-frying and deep-frying. The results determined that the largest mass of PM_{2.5} was generated by deep-frying, followed by stir-frying, boiling and steaming, with average concentrations of PM_{2.5} at 190, 120, 81 and 66 $\mu\text{g m}^{-3}$, respectively. This is consistent with the order of the emission rank of PM_{2.5} in the present study, but the value of PM_{2.5} concentrations were 8 to 13 times higher in the study by See and Balasubramanian³⁰ than the results from our study. Similar experiments were conducted by Wu *et al.*⁵³ with 200 g of bean curd cooked with these same cooking methods. It was found that the most ultra-fine particles (UFPs) were emitted by stir-frying, followed by pan-frying, boiling and steaming, and their particle number emission rates were 1483×10^{10} , 1044×10^{10} , 53×10^{10} and $47 \times 10^{10} \text{ min}^{-1}$, respectively, *i.e.* the emission rate of UFPs from stir-frying was more than 22 times higher than those from boiling and steaming. In the present study, the emission rates of PM_{2.5} from stir-frying was the highest, followed by deep-frying, boiling and steaming. The average PM_{2.5} emission rate of stir-frying was approximately three times greater than that of boiling and steaming. The maximum concentrations and emission rates of boiling ($0.47 \times 10^6 \# \text{ cm}^{-3}$; $0.888 \times 10^{10} \# \text{ s}^{-1}$) and steaming ($0.36 \times 10^6 \# \text{ cm}^{-3}$; $0.783 \times 10^{10} \# \text{ s}^{-1}$) were found to be similar, which could be interpreted by a slower growth in the concentrations of UFPs, as the residual oil and food attached to the pot and hob were heated as well, which could undergo pyrolysis.⁵³ The results of PM mass concentrations corresponding to the methods of cooking, heating sources and the use of extractors are displayed in Table 4 also in comparison with other studies.

Emission rates of PM₁₀ and PM_{2.5} from different cooking methods with and without the extractors in the five study kitchens on are displayed in Table 5. While we found the same order for the PM₁₀ emission rates in terms of cooking methods as for the concentrations of PM₁₀ (see Table 2), for PM_{2.5}, the highest median emission rate was produced by stir-frying with the median PM_{2.5} emission rate of deep-frying only being the second highest. The highest emission rates of both PM₁₀ and PM_{2.5} at a single timepoint were found to be generated by deep-frying with values of 7586 $\mu\text{g min}^{-1}$ and 1229 $\mu\text{g min}^{-1}$, respectively. Emission rates of PM from cooking are impacted by the type of appliance used, the condition of cooking, temperatures and the fat content of the ingredients.⁵⁸ Our results in Table 5 showed that deep-frying and stir-frying had much higher emission rates than boiling and steaming, in line with several previous studies; larger particle production is caused by cooking with oil as opposed to water.^{30,50} Additionally, the type of ingredients used affects emission rates due to the



Table 4 Mass concentrations of particulate matter influenced by cooking methods, heating sources and use of extractors

Cooking method	Heating source	Range hood mode	Average mass concentration ($\mu\text{g m}^{-3}$)		Reference
			PM ₁₀	PM _{2.5}	
Boiling	Gas stove	Off		241.6	Zhao <i>et al.</i> ⁵⁴
Pan-fry				214.1	
Stir-fry				287.5	
Boiling				252.3	
Boiling	Gas stove	On	105.3	81.1	Lee <i>et al.</i> ³¹
Steaming			33.9	28.7	
Deep-fry				230.9	
Stir-fry	Electric hob	On		281	Zhang <i>et al.</i> ⁴³
Stir-fry	Gas stove	On		155	Wong <i>et al.</i> ⁵⁵
Stir-fry	Electric hob			155	
Mix	Gas stove		30.3	24.5	Pan <i>et al.</i> ⁵⁶
Mix				74.9	
Mix	Gas stove	On		75.9	Wan <i>et al.</i> ⁵⁰
Mixed	Gas stove		312.4		See and Balasubramanian ³⁰
Pan-fry	Gas stove	On	1020		To and Yeung ⁵²
Pan-fry	Electric hob		520		
Stir-fry	Gas stove		1330		
Stir-fry	Induction hob		1030		
Deep-fry	Gas stove		4720		
Deep-fry	Induction hob		3980		
Boiling (stew)	Electric hob		117	26.1	Embiale <i>et al.</i> ⁵⁷
	Charcoal		179	50.2	
	Kerosene		124	31.4	
Deep-fry	Electric stove	On	154.3	21.8	Current study
Stir-fry			68.7	11.5	
Boiling			20.5	6.2	
Steaming			13.9	5.0	

Table 5 Ranges and medians of mean PM₁₀ and PM_{2.5} emission rates (ER, $\mu\text{g min}^{-1}$) of cooking activities of each cooking method with or without extractors on (range (median))

	Extractor off		Extractor on	
	ER _{PM₁₀}	ER _{PM_{2.5}}	ER _{PM₁₀}	ER _{PM_{2.5}}
Deep-frying	636–1015 (921)	16–83 (50)	86–913 (535)	17–221 (58)
Stir-frying	303–1720 (546)	18–399 (54)	39–1163 (268)	4–64 (17)
Boiling	39–420 (124)	2–54 (16)	37–139 (119)	6–28 (16)
Steaming	30–109 (72)	4–32 (13)	41–98 (64)	8–15 (14)

percentage of fat contained: food with a high proportion of fat tends to result in higher emission rates than ingredients with low fat proportions.²⁹ In terms of the processes of oil heating, the emission rates varied considerably as well. It was reported that the range of emission rates of PM₁₀ was between 0.67 and 2.33 mg s⁻¹, while the range of PM_{2.5} emission rates varied from 0.06 to 1.46 mg s⁻¹.^{23,59}

The effects of air quality intervention using kitchen extractors

According to Khalid and Foulds⁶⁰ from the UK Energy Research Centre, approximately 60% of the UK residential and commercial kitchens prefer to use natural gas for cooking. Previous research has shown that natural ventilation may not be able to provide a sufficiently high air exchange rate to move indoor aerosol particles outdoors.⁶¹ In Table 5, PM emission rates with

and without the extractors are displayed. Significant effects (at 99.9% confidence level) from the use of extractors were observed for the emission rates of PM₁₀ from deep-frying and stir-frying. The extractors worked well, with removal efficiencies for PM₁₀ and PM_{2.5} of 21.9% to 53.5% and 10.5% to 63.0%, respectively (see Fig. S5†). The highest removal efficiencies for both PM₁₀ and PM_{2.5} were found for stir-frying, followed by deep-frying and steaming. The average removal efficiency of PM₁₀ from oil-based cooking methods in the present work was 51.9%, which was similar to the results from Kang *et al.*⁶² at 45.2% (see Table 6). The removal efficiency of PM_{2.5} from stir-frying in our study was 10.5%, which was much lower than from deep-frying (63.0%), which fitted well with results from Xu and Huang.⁶³ Even though the water-based cooking methods did not contribute large PM emission rates compared with oil-based cooking methods, the extractors still performed with high efficiencies on clearing particles at the average values of 26.4% and 29.1% on PM₁₀ and PM_{2.5}, respectively.

PM emissions of the four typical cooking methods with the extractor on and off are displayed in Fig. 2. Each cooking episode is divided into four stages: (i) is the background testing, (ii) is heating of the pot or pan and the oil or water, (iii) represents the main cooking processes and (iv) is the post-cooking decay after the end of the entire cooking process. The average background concentrations of PM₁₀ and PM_{2.5} for all examined cooking episodes were $4.6 \pm 1.8 \mu\text{g m}^{-3}$ and $2.3 \pm 1.7 \mu\text{g m}^{-3}$, respectively. For deep-frying, with no kitchen extractor, the PM



Table 6 Results of particulate matter removal efficiencies from previous related studies

Cooking method	Volume of room (m ³)	Range hoods exhaust air volume rate (m ³ min ⁻¹)	Air exchange rate (min ⁻¹)	Total removal rate (min ⁻¹)		Removal efficiency		Reference
				PM ₁₀	PM _{2.5}	PM ₁₀	PM _{2.5}	
Mixed	Average 256.57	N/A	0.077	0.0941	0.079	45.24%	49.68%	Kang <i>et al.</i> ⁶²
Mixed	10.88	15	0.36				58%	Chen <i>et al.</i> ⁴⁰
Stir-fry	N/A						59.61%	Xu and Huang ⁶³
Stir-fry							14.29%	
Deep-fry	Average 43.70	7.5	Average 0.075	0.069	0.080	53.53%	62.96%	This study
		Off		0.066	0.079			
Stir-fry		7.5		0.128	0.092	50.25%	10.54%	
		Off		0.123	0.079			
Boiling		7.5		0.172	0.161	21.90%	14.12%	
		Off		0.151	0.143			
Steaming		7.5		0.192	0.140	30.94%	44.04%	
		Off		0.180	0.068			

concentrations rose steadily while heating the oil, and PM_{2.5} (79.0 µg m⁻³) peaked when the chicken was put in with the oil at a high temperature, while PM₁₀ peaked two minutes later than PM_{2.5} with the concentration at 274.7 µg m⁻³ which was 91.5 times above background concentrations. A similar trend was found with the extractor on with higher concentration peaks for PM_{2.5} (81.5 µg m⁻³) and PM₁₀ (338.8 µg m⁻³). When the deep-frying finished, the concentrations of PM₁₀ with extractor fans on and off were 149.4 µg m⁻³ and 254.5 µg m⁻³, respectively. In terms of stir-frying, the PM_{2.5} concentrations under both conditions peaked (extractor on: 62.4 µg m⁻³; off: 73.4 µg m⁻³) at the second minute when the chicken was placed into the pan. When the extractor was on, PM₁₀ concentration (257.4 µg m⁻³) peaked one minute later than the peak of PM_{2.5}. However, with the extractor off, there was an initial PM₁₀ peak (268.4 µg m⁻³) at the same time as the PM_{2.5} peak, with the highest PM₁₀ peak (287.7 µg m⁻³) occurring three minutes later. Under both conditions, PM₁₀ and PM_{2.5} concentrations waned after their peak. Not all periods of post-cooking decay were recorded until the concentrations fully descended back to background levels. When the kitchen extractors were working, both average and peak concentrations of PM₁₀ and PM_{2.5} were significantly lower than those with the extractors off.

For PM₁₀, there were second peaks appearing after a first peak with intervals of 1 to 3 minutes. This could be due to coagulation of UFPs and gaseous pollutants such as VOCs forming secondary organic aerosol (SOA). With the constant emissions of primary organic aerosol (POA), these SOA may cause increasing particle numbers, even though the decay and removal occurred simultaneously. After the second peak, the trend of PM concentrations either remained at relatively low levels or started to decrease steadily. When intervening with the extractors, the slopes of the decays were much steeper, which revealed a swifter decay and removal of the particles indoors, even if it was still within the cooking processes. This suggests that it may not be sufficient to rely on uncontrolled natural ventilation for the removal of the particles. A relatively stable supply of air *via* an open kitchen door, fresh air systems and

kitchen extractors could work much more effectively together to reduce the particles in the breathing zone, rather than only opening the kitchen window.^{23,50} In addition to the volume flow rate of the extractors and air exchange rates, there are other factors, such as the volume of the kitchen and the total removal rates, including the decay rate, that should be taken into account. The summary of comparisons of our work to other studies regarding the efficiencies of extractors operation is presented in Table 6.

PM residence times with extractors on and off. Residence time of the PM generated by cooking was counted from the time of the end of each cooking activity until the concentration of PM returned to the background level. Deep-frying and stir-frying could emit PM₁₀ in high quantities, after which the PM concentrations decreased to below the limit given by the World Health Organization (24 hour mean PM₁₀: 45 µg m⁻³);⁴⁷ the time taken for PM concentrations to return to a safe level was also measured.

Table 7 presents the durations of post-cooking decay for each cooking method and the time taken for PM₁₀ to decrease to the WHO limit. With the extractor on, the mean duration for PM₁₀ and PM_{2.5} generated by deep-frying and stir-frying to return to the background level was significantly shorter (PM₁₀: 27.1–38.8%, PM_{2.5}: 26.5–29.3%) than with the extractors off, with confidence levels at 99.9%. For boiling and steaming, the use of extractors also sped up (by 32.0% to 40.5%) the reduction of PM₁₀ significantly at 99% confidence levels. However, no significant effects of extractors on the removal of PM_{2.5} during post-cooking decay were found, although the average durations were shortened. In terms of the reductions of PM₁₀ to the recommended WHO limit, when the extractors were off, it took approximately a third of the entire duration of PM₁₀ post-cooking decay to reduce PM₁₀ concentrations lower than 50 µg m⁻³ for both emissions by deep-frying and stir-frying. It was significantly (by 47.1% to 75.5%) faster to reduce PM₁₀ emitted by deep-frying and stir-frying below the WHO level of health concern by using the kitchen extractor at a 99.9% confidence level.



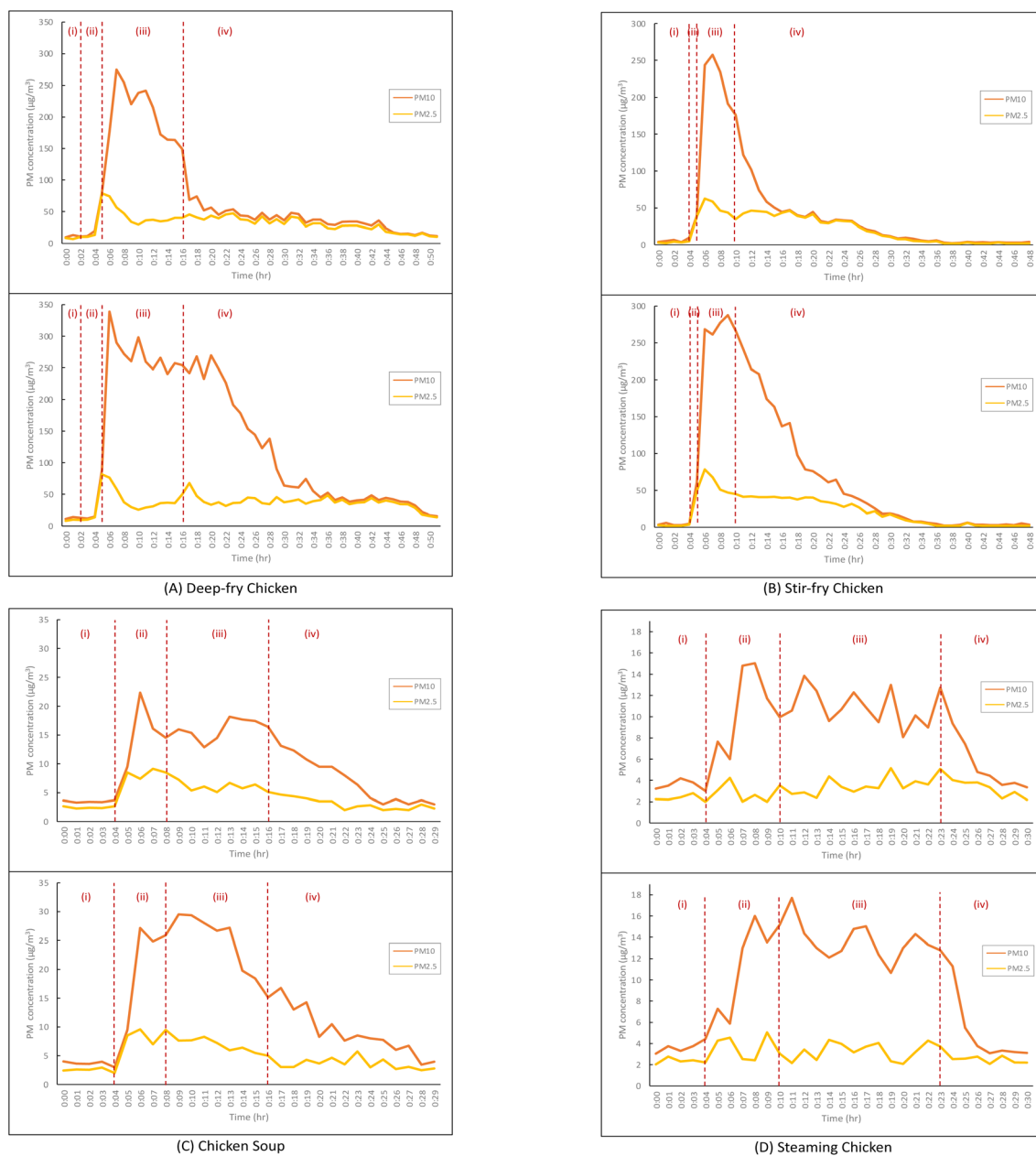


Fig. 2 Particulate matter emissions during the cooking process of four cooking episode. (A) Deep-fry chicken, (B) stir-fry chicken, (C) chicken soup and (D) steaming chicken, with extractors (upper image) on and (lower image) off. Four stages of cooking process: (i) the background level; (ii) heating of the pot or pan and the oil or water (iii) the main cooking processes; (iv) the post-cooking decay after the end of the entire cooking process.

Table 7 Residence time (the duration when hobs were turned off to the pollutant concentrations returning to the background level) of PM_{10} , $PM_{2.5}$ and the length of time for PM_{10} to be reduced to below the WHO limit (24 hour mean PM_{10} : $45 \mu g m^{-3}$) with the extractors on and off (mean duration \pm standard deviation) (min)

	PM_{10}		$PM_{2.5}$		PM_{10} (WHO standard)	
	Extractors off	Extractors on	Extractors off	Extractors on	Extractors off	Extractors on
Deep-fry	53.0 ± 9.8	32.4 ± 2.1	20.8 ± 2.8	14.7 ± 2.2	23.2 ± 3.7	6.4 ± 1.5
Stir-fry	31.6 ± 2.0	23.1 ± 2.3	17.6 ± 3.8	12.9 ± 5.2	11.6 ± 2.1	6.2 ± 1.4
Boiling	9.2 ± 2.5	5.5 ± 1.6	6.2 ± 2.2	4.3 ± 1.8	—	—
Steaming	10.0 ± 1.8	6.8 ± 1.9	4.1 ± 1.6	3.4 ± 1.1	—	—





Fig. 3 (A) PM_{10} and (B) $PM_{2.5}$ concentrations with the extractors on/off during post-cooking decay with exponential regressions.

The PM pollutant is presented in either the air compartment or on a surface compartment.⁶⁴ In the present study, the surface compartment residence time was ignored as it measured the average time a particle remained on a surface before being removed by cleaning or resuspension, while the air compartment residence time examined the average time spent to remove a particle in the indoor air *via* deposition and exfiltration.⁶⁴ The post-activity decreasing trends of particles of all sizes were found to be following an exponential decay at the room air exchange rate for this study and in line with previous work.⁶⁵ It should be noted that the air compartment residence time of PM is impacted by the total removal rate (λ) which includes the air exchange rate and the coagulation and deposition of the particles.

With no mechanical ventilation presented, after a cooking event, the concentrations of nanosized particles larger than 6 nm (*ca.* 6–1000 nm) ($PN_{>6}$) started to decay due to the mixing of pollutants throughout the room, or by being removed by ventilation, infiltration and deposition.^{65,66} In the current study, the air exchange rate was assumed to be 0.075 min^{-1} (4.5 h^{-1}) based on the studies by Kang *et al.*⁶² and Lunden *et al.*⁶⁷ Lunden *et al.*⁶⁷ conducted two experiments of pan-frying and stir-frying, with the results of the residence time of $PN_{>6}$ being 19 and 26 minutes, respectively, while their peak concentrations were approximately 10^4 and $10^5 \# \text{ cm}^{-3}$. Due to its similar conditions of air exchange rates and volume of kitchen, the residence times observed by Lunden *et al.*⁶⁷ fit into the range of the current study. However, because of a low air exchange rate, much longer residence times of particles were found by Singer *et al.*⁶⁶ who performed repeated experiments investigating cooking emissions of $PN_{>6}$ and gaseous pollutants. $PN_{>6}$ were found to deposit indoors at a rate which was quick enough to compete against the air exchange as a removal process, and the average time of decay was approximately 88 minutes with an air exchange rate of 0.0083 min^{-1} .⁶⁶ Similarly, in the research by Qian *et al.*,⁶⁴ the air exchange rates ranged between 0.11 – 0.58 h^{-1} (0.0018 – 0.0097 min^{-1}), which were significantly lower, and the average residence time of PM_{10} and polystyrene latex (PSL)

tracer particles were in 1.2 and 2.6 hours, respectively, in the air compartment.

The average post-cooking PM decays were fitted with regression lines to exponential functions as illustrated in Fig. 3. The averaged PM concentrations after cooking at each time point are displayed together with the regression lines. The regression trend lines generally fit well the decay of PM_{10} and $PM_{2.5}$ with and without the kitchen extractors on after cooking activities as demonstrated by high correlation coefficients (all the R^2 values are greater than 80%). Although the PM_{10} concentrations did not quite reach the background levels when the cooking activities finished, with the extractors on, PM_{10} concentrations dropped 1.43 times faster and the slope of the first three minutes was much steeper than without extractors. The effect of extractors on the post-cooking decay of $PM_{2.5}$ concentration was not as clear, but a slightly steeper slope of the $PM_{2.5}$ decay with kitchen extractors on was observed in the first five minutes.

During the PM decay in the present study, the rates of particle removal were higher for PM_{10} than for $PM_{2.5}$ as shown in Fig. 3. This was because, besides the air exchange rate, particle deposition rates significantly contributed to the PM removal rate. This varied broadly across the various conditions, including particle size variation, which was reported to be one of the most critical factors impacting on indoor particle concentrations and removal rates, together with the quantity of interior furnishings.^{68,69} Thatcher *et al.*⁷⁰ conducted experiments to explore the deposition rates of particles by using three different levels of indoor furnishings and four different conditions of air flow in an isolated room with a volume of 14.2 m^3 . With an increase of the particle diameter from 0.55 to $8.66 \mu\text{m}$, the deposition loss rate coefficients increased from 0.10 to 6.79 h^{-1} (0.0016 – 0.1132 min^{-1}) in an unfurnished room with a bare, electrically grounded metal floor. For an unfurnished room with a carpeted floor and one that was fully furnished, the deposition loss rate coefficients increased together with the particle diameter (0.55 – $8.66 \mu\text{m}$), varied 0.12 – 4.92 h^{-1} (0.002 – 0.082 min^{-1}) and 0.20 – 5.54 h^{-1} (0.0033 – 0.908 min^{-1}),



respectively.⁷⁰ This indicated that the residence time of the particles with the diameter from 0.55 to 8.66 μm ranged from approximately 10 hours to only 8.83 minutes, and the residence time for particles with a mean diameter at 2.37 μm was about 65 minutes.⁶⁴ In our present study, the trends of PM_{10} and $\text{PM}_{2.5}$ decays showed that PM_{10} had a dramatically faster decay rate in the first 14 minutes compared to that of $\text{PM}_{2.5}$, which indicates that the PM_{10} deposition occurred much swifter than for $\text{PM}_{2.5}$.

The use of kitchen extractors significantly raised the efficiency of the PM removal and reduced the PM residence time. Quicker depositions were found in the first three minutes after the peak concentrations of PM_{10} when the extractors were on compared to those with extractors off. From Table 7, it is clear that the residence time of both PM_{10} and $\text{PM}_{2.5}$ decreased due to the use of extractors. However, the performance of extractors on $\text{PM}_{2.5}$ removal was not as clear as no statistical significance was found between the groups with and without extractors on, which might be because of their relatively lower concentrations at the peak time. Singer *et al.*⁶⁶ determined that extractors performed with higher reduction efficiency on PN with a larger diameter than those in small sizes, resulting in approximately 130% faster decay. A similar result was found by Rim *et al.*⁷¹ reporting that removal effectiveness for particles with diameters of 2–6 nm (PN_{2-6}) was lower than for particles larger than 6 nm ($\text{PN}_{>6}$), which led to approximately a 59% shorter time for $\text{PN}_{>6}$ to decay than for PN_{2-6} . In Table 7, it also indicates that an additional five-minute running of the extractors could reduce PM_{10} to a safe level after deep-frying and stir-frying.

Influences of oil type and temperature

As the oil-based cooking methods contributed higher PM emissions than water-based methods, the type of oil could also play a critical role in the generation of aerosols. It was shown that the composition of oil can influence the temperature at which the oil started to decompose and emit visible smoke fumes.¹⁹ When the semi-volatile compounds are released during the heating of oil, they tended to be condensed and form liquid aerosols, which eventually generated particles.²⁹ Thus, the PM emission rates are influenced by the proportions of semi-volatile and volatile species in the oil. According to Gao *et al.*²³ and Torkmahalleh *et al.*,⁵⁹ the ranges of concentrations and emission rates of PM_{10} were 7.4–30 mg m^{-3} and 0.51–2.33 mg s^{-1} , respectively, while those of $\text{PM}_{2.5}$ were 6.5–18.8 mg m^{-3} and 0.05–1.46 mg s^{-1} respectively, due to different oil types. Based on the study by Liu *et al.*,¹⁴ POA ranged from 0.8 to 42.3 $\mu\text{g m}^{-3}$, with olive oil emitting the highest POA, and SOA ranged from 27.1 to 107.5 $\mu\text{g m}^{-3}$, with palm oil generating the greatest quantity of SOA. Rapeseed oil, peanut oil and olive oil generated a relatively high particle number, while soybean oil, sunflower oil and blend oil released the lowest concentrations.^{14,23} In general, oils with higher smoke points tend to generate lower concentrations of particles.¹⁹ It was found that when the temperature of oil was between 180 and 210 $^{\circ}\text{C}$, every five-degree increase could lead to a rise in the concentrations of UFPs of 20–50%.⁷² The mode diameter of oil droplets was found

to grow with increasing temperature, leading to enhanced concentrations of accumulation mode aerosols.⁷²

Personal exposure to cooking PM and health effects

Personal exposure assessment. In the present study, the average cooking durations of deep-frying, stir-frying, boiling and steaming were 7.8, 9.4, 29.7 and 17.9 minutes, respectively. As the studied student accommodation were studio flats, inhabitants would be exposed to pollutants after cooking activities as well, so the post-cooking decay durations were also taken into account. The average exposure durations for each cooking method are displayed in Table 8. According to the surveys, the average number of dishes per cooking activities (\bar{N}) was 1.37, and the averaged frequency of cooking activities (f_c) was 1.94 times per day. In Table 3, the exposure time of the four cooking methods is presented. The total exposure time per day (t_e) for PM_{10} and $\text{PM}_{2.5}$ was calculated by using eqn (13) resulting in 103.81 and 77.09 minutes, respectively.

$$t_e = \sum_{i=1}^N (t_1 - t_0) / N \times 1.37 \times 1.94 \quad (13)$$

The average exposure time of each cooking method per day was calculated by multiplying t_e by the *likelihood of choosing a particular cooking method per day*, listed in the 4th column of Table 8.

The annual intake for PM_{10} and $\text{PM}_{2.5}$ are displayed in Table 9. The PM_{10} annual intake ranked with the same order as the PM_{10} concentrations, even though deep-frying only contributed 12% to the four cooking methods. Steaming contributed 14% to the cooking activities but its PM_{10} annual intake was about 18 and 14 times lower than that of deep-frying and stir-frying. Boiling had the largest share in cooking activities, though it donated only $\frac{1}{3}$ and $\frac{2}{5}$ of deep-frying and stir-frying annual inhalation exposures, respectively. In terms of the annual $\text{PM}_{2.5}$ intake, it was noticed that boiling played the greatest role, followed by stir-frying and deep-frying, with steaming the lowest, which was about 5.7 times lower than that of boiling. The PM_{10} and $\text{PM}_{2.5}$ intake of male operators in the student flats were approximately 12% higher than those of females, as the average inhalation rates used for calculation were 0.66 $\text{m}^3 \text{h}^{-1}$ and 0.59 $\text{m}^3 \text{h}^{-1}$ for males and females, respectively.

According to Table 9, the average annual intake to PM_{10} and $\text{PM}_{2.5}$ from cooking emissions in the studio flats were 18.64 and 2.27 mg per year, respectively. Deep-frying and stir-frying produced high values of particles, even though their cooking durations were the shortest. They also took much longer to decay to the background levels, resulted in longer exposure duration, as shown in Table 7. The exposure by deep-frying contributed the largest fraction of PM_{10} while it took the second lowest proportion of average cooking time per day due to its relatively short cooking duration and low number of times chosen. Stir-frying and boiling were two of the most popular cooking methods, but the PM_{10} intake of stir-frying was nearly three times higher than that of boiling, because of its high emission rate. Also, stir-frying contributed the highest $\text{PM}_{2.5}$



Table 8 Exposure time from each cooking method per day and proportion of the cooking method frequency people chose over the entire surveyed period

	Average exposure duration per dish (min)		Likelihood of choosing a particular cooking method per day	Average exposure time of each cooking method per day (min)	
	PM ₁₀	PM _{2.5}		PM ₁₀	PM _{2.5}
Deep-frying	48.8	25.0	12%	16.0	8.2
Stir-frying	40.5	26.3	33%	35.5	23.1
Boiling	39.3	35.0	40%	42.0	37.4
Steaming	26.9	21.8	14%	10.3	8.4

Table 9 Annual particulate matter intake (D_{pot}) (μg per year)

	D_{pot}		D_{pot} for male		D_{pot} for female	
	PM ₁₀	PM _{2.5}	PM ₁₀	PM _{2.5}	PM ₁₀	PM _{2.5}
Deep-fry	7789	564	8160	591	7294	528
Stir-fry	7682	836	8048	876	7195	783
Boiling	2713	736	2842	771	2540	689
Steaming	452	132	473	138	423	124
Total	18 636	2268	19 523	2376	17 452	2124

intake because of its high frequency and oil-based method. Due to the low emission rate and short cooking duration, steaming provided particularly low intake.

Liao *et al.*⁵¹ studied the PM mass lung/indoor (L/I) ratio and the inhalation exposure dose of major compartments in the human respiratory tract by using a L/I ratio model. It was suggested that the integrated cumulative inhalation dose rates (particles per cm² per h) in the nasal passage (23.94–24.27), bronchial region (4.92–5.06), bronchiolar region (4.97–10.82) were significantly higher than those in the alveolar interstitial region (0.002–0.02) for the events of cooking. In terms of the L/I ratios of PM mass, the regions of nasal passage and pharynx (0.70–0.83) were found to be higher than those of the bronchial region (0.41–0.62), bronchiolar region (0.12–0.41), alveolar interstitial region (0.02–0.26), where they discovered larger sized particles (diameter >3 μm) with smaller L/I ratios of PM.⁵¹ This explained that, due to the deposition during respiration, particles, especially those with bigger size ranges, tended to be no longer airborne, resulting in a lower concentration of PM in deeper lung regions. The greater extent of harmful particles in the lung were those in smaller size ranges (diameter < 3 μm).

Lu *et al.*⁸ had conducted experiments on assessing the personal inhalation exposure to PM_{2.5} during Chinese family cooking. The results showed that the exposure of male and female operators was 346.3 and 309.6 mg per year, respectively, due to physiological differences, which resulting in various amounts of inhalation while breathing. In the present study, the differences were identified as well, but the results were approximately 150 times lower than those from the study by Lu *et al.*⁸ According to Ji *et al.*,⁷³ the annual inhalation exposure of indoor PM_{2.5} from daily life was 174.84 mg per year, which was approximately 10 times higher than the present study's

findings. According to Mohammadyan,⁷⁴ the concentrations of indoor PM were the best predictors of personal exposure. The measured average concentrations of PM_{2.5} were the major difference, which were 60 $\mu\text{g m}^{-3}$ by Ji *et al.*⁷³ and 599 $\mu\text{g m}^{-3}$ by Lu *et al.*⁸ as they used gas stoves, while the present study result was 9.36 $\mu\text{g m}^{-3}$ using electric hobs. The experiments by Lu *et al.*⁸ counted 3.54 dishes per meal, two times of cooking activities per day, and 365 days of the exposure frequency as it was designed for a family with four people comprising three adults and one child, while the values were 1.37, 1.94 and 300 respectively in the present study for only one adult student. This underlines that the type of stove and duration of cooking plays a critical role in the personal exposure. In addition, the background level of air quality needs to be taken into account. The outdoor concentrations of PM_{2.5} in these two previous studies ranged from 10 to 220 $\mu\text{g m}^{-3}$; especially the study by Lu *et al.*⁸ which took place in Tianjin, an industrial city in Northern China, during the late winter, was during a period with relatively serious air pollution, resulting in a background PM_{2.5} range of 41 to 168 $\mu\text{g m}^{-3}$. In contrast, the data collection period of the present study was during the early summer and a national lockdown was implemented due to the COVID-19 pandemic, which resulted in the PM_{2.5} concentration ranging from 2.8 to 6.2 $\mu\text{g m}^{-3}$.

Health effects of cooking PM. Based on the results from Lu *et al.*⁸ and Ji *et al.*,⁷³ the annual inhalation exposures of PM_{2.5} from cooking in ordinary families for males and females were 1.98 and 1.77 times higher than the general population. Additionally, personal exposure to PM_{2.5} was discovered to be higher than the measured PM_{2.5} concentration indoor and outdoor, due to a “personal cloud” effect.^{74,75} This was dependent on the size of particles, with the personal cloud levels of PM₁₀ expected to be six to seven times higher than those of PM_{2.5}.⁷⁶ Epidemiological research confirmed the risk of personal exposure to PM. Indoor cooking oil fume exposure was correlated with an increased risk of respiratory diseases, lung cancer, cardiopulmonary endpoints, and rising levels of urinary 8-OHdG.^{17,56} According to the Health Risk Assessment for Air Pollutants by Golder Associates,⁷⁷ changes of exposure to daily PM_{2.5} and PM₁₀ from 25 and 50 $\mu\text{g m}^{-3}$ (WHO standards),⁷⁸ to 15 and 30 $\mu\text{g m}^{-3}$ could reduce approximately 40% and 50% of the mortality from cardiovascular diseases. An additional 1.2 years could be gained of expected lifespans for people under 40 years old when the annual mean levels of PM_{2.5} were lower than 15 $\mu\text{g m}^{-3}$.⁷⁹



Stir-frying was one of the most popular cooking methods and it produced the highest amount of annual $\text{PM}_{2.5}$ inhalation exposure in the present study. It has been demonstrated by other studies that stir-frying significantly raises the levels of indoor $\text{PM}_{2.5}$, which may impair autonomic function and decrease the heart rate variability indices eventually.⁸⁰

Cooking activities also contribute greatly to people's response to oxidative stress, especially for the vulnerable population.⁸ The relationship between oxidative stress response and the concentration of $\text{PM}_{2.5}$ was assessed by Kim *et al.*,⁸¹ who demonstrated that the acceptable mean personal exposures were 20.7 ± 12.7 and $80.5 \pm 29.9 \mu\text{g m}^{-3}$ for young children and the elderly, respectively, due to different bodily functions. Although, in the present study, the mean concentrations of $\text{PM}_{2.5}$ did not exceed these levels, previous studies have reported much higher personal exposures to indoor $\text{PM}_{2.5}$ due to cooking. A recent systematic review on the indoor air pollution impact on the health of children and people with pre-existing lung disease by Maung *et al.*⁸² reported that $\text{PM}_{2.5}$ levels vary seasonally with the highest levels reached in winter. From these studies it was strongly suggested that vulnerable people should carefully consider their participation in cooking processes and accessible intervention options.

Limitations of the study

Deploying low-cost sensors in real-world scenarios has the potential to deliver important insights to complement more rigorous, but also more costly as well as time- and location-limited studies with reference instruments. Apart from the limitations in the accuracy and precision of the low-cost sensors themselves (a comparison with reference instruments *e.g.* through co-location is essential to establish the performance for each individual pollutant of interest), there are a number of further uncertainties that need to be taken into account: air mixing regimes at the chosen sensor locations, variations in building ventilation caused *e.g.* by local wind conditions, building fabrics, building state and maintenance, as well as influences from external factors such as neighbouring flats as pollution sources and variable outdoor pollutant ingress into the study location are a few examples of such uncertainties. However, the most important uncertainties are likely associated with the behaviours of occupants and the interpretation of pollution measurements largely relies on the accuracy of the activity profiles reported by the volunteers which should ideally not only include the cooking activities, but also any activity that affects the indoor air quality such as opening/closing of windows and doors as well as any other activities that generate or dilute pollutant levels indoors; given the complexity of the activity profiles even for a single occupant, use of increasingly affordable smart technology to reduce the reporting burden on the volunteer could be an important step to reduce the uncertainties associated with this type of study.

Conclusion

The present study identified the factors influencing indoor particulate matter levels from cooking emissions, including the

cooking methods and the impact of extractors. In addition, the residence time and personal exposure to the cooking-generated PM was measured using a low-cost sensor following validation of the PM data by co-location of the sensor with a reference instrument. Oil-based cooking methods generated much higher PM levels than water-based methods. The use of kitchen extractors provided effective ventilation removing the pollutants, lowering the emission rates and speeding up the pollutant decay. This all resulted in a shorter residence time of the particles indoors. When considering the frequency of each cooking method, the personal exposure varied widely. It is recommended to minimise exposure to PM cooking emissions by choosing to cook water-based dishes, turning on the extractor for a straightforward indoor air quality intervention and leaving the extractor on for an additional five to ten minutes once the cooking process has finished. Although the accuracy of the low-cost sensor has scope for improvement, it successfully determined the trends of PM concentrations from the activities studied. This enables the public to raise their awareness of indoor air pollution caused by cooking activities and seek to apply strategies to maintain an all-round healthier lifestyle. Further research should focus on more effective ways to remove the pollutants generated in the various cooking activities to reduce personal exposure and health impacts.

Author contributions

Conceptualisation was jointly by CP & RT; funding acquisition, supervision and project administration was carried out by CP; investigation, visualisation, data curation, formal analysis, validation and writing the original draft of the manuscript was carried out by RT; reviewing and editing of the manuscript was performed jointly by CP & RT.

Conflicts of interest

There are no conflicts to declare.

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References

- 1 D. Norbäck, in *Indoor Environmental Quality and Health Risk toward Healthier Environment for All*, 2020, ch. 17, pp. 321–333, DOI: [10.1007/978-981-32-9182-9_17](https://doi.org/10.1007/978-981-32-9182-9_17).
- 2 C. P. Weisel, S. Alimokhtari and P. F. Sanders, *Environ. Sci. Technol.*, 2008, **42**, 8231–8238.
- 3 R. Kamens, C.-t. Lee, R. Wiener and D. Leith, *Atmos. Environ., Part A*, 1991, **25**, 939–948.
- 4 W. G. Kreyling, M. Semmler-Behnke and W. Möller, *J. Nanopart. Res.*, 2006, **8**, 543–562.
- 5 Y. Huang, S. S. H. Ho, K. F. Ho, S. C. Lee, J. Z. Yu and P. K. K. Louie, *J. Hazard. Mater.*, 2011, **186**, 344–351.
- 6 M. Dennekamp, *Occup. Environ. Med.*, 2001, **58**, 511–516.
- 7 K. L. Abdullahi, J. M. Delgado-Saborit and R. M. Harrison, *Atmos. Environ.*, 2013, **71**, 260–294.
- 8 F. Lu, B. Shen, P. Yuan, S. Li, Y. Sun and X. Mei, *Sci. Total Environ.*, 2019, **654**, 671–677.
- 9 J. A. Adeniran, J. A. Sonibare and L. A. Jimoda, *Atmos. Pollut. Res.*, 2015, **6**, 21–28.
- 10 J. A. Adeniran, R. O. Yusuf, S. I. Mustapha and J. A. Sonibare, *Environ. Res. Eng. Manag.*, 2018, **73**(4), 21–30.
- 11 E. Géhin, O. Ramalho and S. Kirchner, *Atmos. Environ.*, 2008, **42**, 8341–8352.
- 12 T.-T. Yang, S.-T. Lin, R.-H. Shie, C.-K. Tseng and C.-H. Ku, *Aerosol Air Qual. Res.*, 2016, **16**, 2570–2580.
- 13 S. Orecchio, *Atmos. Environ.*, 2011, **45**, 1888–1895.
- 14 T. Liu, Z. Wang, D. D. Huang, X. Wang and C. K. Chan, *Environ. Sci. Technol. Lett.*, 2017, **5**, 32–37.
- 15 E. Abt, H. H. Suh, G. Allen and P. Koutrakis, *Environ. Health Perspect.*, 2000, **108**, 35–44.
- 16 E. Kabir and K.-H. Kim, *J. Hazard. Mater.*, 2011, **188**, 443–454.
- 17 L. Wang, Z. Xiang, S. Stevanovic, Z. Ristovski, F. Salimi, J. Gao, H. Wang and L. Li, *Sci. Total Environ.*, 2017, **589**, 173–181.
- 18 Y. Zhao, C. Chen and B. Zhao, *Atmos. Environ.*, 2018, **193**, 190–197.
- 19 Y. Zhao and B. Zhao, *Build. Simul.*, 2018, **11**, 977–995.
- 20 R. M. Healy, J. Sciare, L. Poulain, M. Crippa, A. Wiedensohler, A. S. H. Prévôt, U. Baltensperger, R. Sarda-Estève, M. L. McGuire, C. H. Jeong, E. McGillicuddy, I. P. O'Connor, J. R. Sodeau, G. J. Evans and J. C. Wenger, *Atmos. Chem. Phys.*, 2013, **13**, 9479–9496.
- 21 L. A. Wallace, S. J. Emmerich and C. Howard-Reed, *Environ. Sci. Technol.*, 2004, **38**, 2304–2311.
- 22 F. Klein, S. M. Platt, N. J. Farren, A. Detournay, E. A. Bruns, C. Bozzetti, K. R. Daellenbach, D. Kilic, N. K. Kumar, S. M. Pieber, J. G. Slowik, B. Temime-Roussel, N. Marchand, J. F. Hamilton, U. Baltensperger, A. S. H. Prévôt and I. El Haddad, *Environ. Sci. Technol.*, 2016, **50**, 1243–1250.
- 23 J. Gao, C. Cao, L. Wang, T. Song, X. Zhou, J. Yang and X. Zhang, *Aerosol Air Qual. Res.*, 2013, **13**, 488–496.
- 24 J. A. Adeniran, R. O. Yusuf, M. O. Abdulkadir, M.-N. O. Yusuf, K. A. Abdulraheem, B. K. Adeoye, J. A. Sonibare and M. Du, *Environ. Pollut.*, 2020, **266**(Pt 2), 115169.
- 25 D. von Bornstädt, A. Kunz and M. Endres, *J. Cereb. Blood Flow Metab.*, 2013, **34**, 215–220.
- 26 W.-T. Tsai, *Environments*, 2016, **3**(4), 23.
- 27 S. Gorjinezhad, A. Kerimray, M. A. Torkmahalleh, M. Keleş, F. Ozturk and P. K. Hopke, *Environ. Sci. Pollut. Res.*, 2017, **24**, 9515–9529.
- 28 J. J. Schauer, M. J. Kleeman, G. R. Cass and B. R. T. Simoneit, *Environ. Sci. Technol.*, 2001, **36**, 567–575.
- 29 G. Buonanno, L. Morawska and L. Stabile, *Atmos. Environ.*, 2009, **43**, 3235–3242.
- 30 S. W. See and R. Balasubramanian, *Environ. Res.*, 2006, **102**, 197–204.
- 31 S. C. Lee, W.-M. Li and L. Yin Chan, *Sci. Total Environ.*, 2001, **279**, 181–193.
- 32 Y. Li and X. Li, *Int. J. Vent.*, 2016, **5**, 273.
- 33 Y. Chen, K. F. Ho, S. S. H. Ho, W. K. Ho, S. C. Lee, J. Z. Yu and E. H. L. Sit, *J. Environ. Monit.*, 2007, **9**(12), 1402–1409.
- 34 Y. Zhao, M. Hu, S. Slanina and Y. Zhang, *Environ. Sci. Technol.*, 2006, **41**, 99–105.
- 35 P. Labs, *Evaluation of Flow, a Personal Air Quality Sensor*, <https://blog.plumelabs.com/2019/06/21/how-accurate-is-flow/>, accessed 04 Oct, 2022.
- 36 N. Crnosija, M. Levy Zamora, A. M. Rule and D. Payne-Sturges, *Int. J. Environ. Res. Public Health*, 2022, **19**(12), 7340.
- 37 F. M. J. Bulot, H. S. Russell, M. Rezaei, M. S. Johnson, S. J. J. Ossont, A. K. R. Morris, P. J. Basford, N. H. C. Easton, G. L. Foster, M. Loxham and S. J. Cox, *Sensors*, 2020, **20**(8), 2219.
- 38 M. Levy Zamora, F. Xiong, D. Gentner, B. Kerkez, J. Kohrman-Glaser and K. Koehler, *Environ. Sci. Technol.*, 2018, **53**, 838–849.
- 39 D. M. Styne and H. McHenry, *Horm. Res.*, 1993, **39**, 3–6.
- 40 C. Chen, Y. Zhao and B. Zhao, *Environ. Sci. Technol.*, 2018, **52**, 1081–1087.
- 41 S. Bhangar, N. A. Mullen, S. V. Hering, N. M. Kreisberg and W. W. Nazaroff, *Indoor Air*, 2011, **21**, 132–144.
- 42 J. Thornburg, D. S. Ensor, C. E. Rodes, P. A. Lawless, L. E. Sparks and R. B. Mosley, *Aerosol Sci. Technol.*, 2001, **34**, 284–296.
- 43 Q. Zhang, R. H. Gangupomu, D. Ramirez and Y. Zhu, *Int. J. Environ. Res. Public Health*, 2010, **7**, 1744–1759.
- 44 Q. Zhang, J. Avalos and Y. Zhu, *Indoor Air*, 2014, **24**, 190–198.
- 45 N. Hodas, M. Loh, H. M. Shin, D. Li, D. Bennett, T. E. McKone, O. Jolliet, C. J. Weschler, M. Jantunen, P. Liroy and P. Fantke, *Indoor Air*, 2015, **26**, 836–856.
- 46 B. Wang and X. Duan, *Epidemiology*, 2011, **22**, S97.
- 47 W. H. Organization, *WHO Global Air Quality Guidelines: Particulate Matter (PM_{2.5} and PM₁₀), Ozone, Nitrogen Dioxide, Sulfur Dioxide and Carbon Monoxide*, World Health Organization, 2021.
- 48 D. f. E. F. R. Affairs, *UK and EU Air Quality Limits*, 2021.
- 49 A. Varlamov, Z. Zhou and Y. Chen, *Phys. Economy*, 2018, **1/2018**, 55–68.



- 50 M.-P. Wan, C.-L. Wu, G.-N. Sze To, T.-C. Chan and C. Y. H. Chao, *Atmos. Environ.*, 2011, **45**, 6141–6148.
- 51 C.-M. Liao, S.-C. Chen, J.-W. Chen and H.-M. Liang, *Sci. Total Environ.*, 2006, **358**, 72–84.
- 52 W. M. To and L. L. Yeung, *Indoor Built Environ.*, 2011, **20**, 555–563.
- 53 C. L. Wu, C. Y. H. Chao, G. N. Sze-To, M. P. Wan and T. C. Chan, *Indoor Built Environ.*, 2011, **21**, 782–796.
- 54 P. Zhao, K.-P. Yu and C.-C. Lin, *Int. Arch. Occup. Environ. Health*, 2010, **84**, 231–237.
- 55 T. W. Wong, A. H. S. Wong, F. S. C. Lee and H. Qiu, *Occup. Environ. Med.*, 2011, **68**, 746–752.
- 56 C.-H. Pan, C.-C. Chan and K.-Y. Wu, *Cancer Epidemiol., Biomarkers Prev.*, 2008, **17**, 3351–3357.
- 57 A. Embiale, F. Zewge, B. S. Chandravanshi and E. Sahle-Demessie, *Indoor Built Environ.*, 2019, **28**, 1140–1154.
- 58 J. D. McDonald, B. Zielinska, E. M. Fujita, J. C. Sagebiel, J. C. Chow and J. G. Watson, *J. Air Waste Manage. Assoc.*, 2012, **53**, 185–194.
- 59 M. A. Torkmahalleh, I. Goldasteh, Y. Zhao, N. M. Udochu, A. Rossner, P. K. Hopke and A. R. Ferro, *Indoor Air*, 2012, **22**, 483–491.
- 60 R. Khalid and C. Foulds, *The Social Dimensions of Moving Away from Gas Cookers and Hobs: Challenges and Opportunities in Transition to Low-Carbon Cooking*, The Department for Business, Energy & Industrial Strategy (BEIS), 2020.
- 61 T. Hussein, T. Glytsos, J. Ondráček, P. Dohányosová, V. Ždímal, K. Hämeri, M. Lazaridis, J. Smolík and M. Kulmala, *Atmos. Environ.*, 2006, **40**, 4285–4307.
- 62 K. Kang, H. Kim, D. D. Kim, Y. G. Lee and T. Kim, *Sci. Total Environ.*, 2019, **668**, 56–66.
- 63 J. Xu and Z. Huang, *J. Xi'an Univ. Posts Telecommun.*, 2014, **19**(5), 101–105.
- 64 J. Qian, A. R. Ferro and K. R. Fowler, *J. Air Waste Manage. Assoc.*, 2012, **58**, 502–516.
- 65 V. Y. Seaman, D. H. Bennett and T. M. Cahill, *Atmos. Environ.*, 2009, **43**, 6199–6204.
- 66 B. C. Singer, R. Z. Pass, W. W. Delp, D. M. Lorenzetti and R. L. Maddalena, *Build Environ.*, 2017, **122**, 215–229.
- 67 M. M. Lunden, W. W. Delp and B. C. Singer, *Indoor Air*, 2015, **25**, 45–58.
- 68 R. B. Mosley, D. J. Greenwell, L. E. Sparks, Z. Guo, W. G. Tucker, R. Fortmann and C. Whitfield, *Aerosol Sci. Technol.*, 2001, **34**, 127–136.
- 69 T. Thatcher and D. Layton, *Atmos. Environ.*, 1995, **29**, 1487–1497.
- 70 T. L. Thatcher, A. C. K. Lai, R. Moreno-Jackson, R. G. Sextro and W. W. Nazaroff, *Atmos. Environ.*, 2002, **36**, 1811–1819.
- 71 D. Rim, L. Wallace, S. Nabinger and A. Persily, *Sci. Total Environ.*, 2012, **432**, 350–356.
- 72 L. Tseng and C. Chen, *Ann. Occup. Hyg.*, 2013, **57**, 920–933.
- 73 H. Ji, Y. Fang, T. Song and C. Cao, *Environ. Pollut. Control*, 2014, **36**(9).
- 74 M. Mohammadyan, *Iran. J. Energy Environ.*, 2012, **3**(3), 246–254.
- 75 L. Wallace, *J. Air Waste Manage. Assoc.*, 1996, **46**, 98–126.
- 76 L. Rojas-Bracho, H. H. Suh, P. J. Catalano and P. Koutrakis, *J. Air Waste Manage. Assoc.*, 2012, **54**, 207–217.
- 77 G. Associates, *Exposure Assessment and Risk Characterisation to Inform Recommendations for Updating Ambient Air Quality Standards for PM_{2.5}, PM₁₀, O₃, NO₂, SO₂* Golden Associates, Canberra, 2013.
- 78 WHO, *WHO Air Quality Guidelines for Particulate Matter, Ozone, Nitrogen Dioxide and Sulfur Dioxide*, World Health Organization, 2015.
- 79 S. Medina, A. Le Tertre and M. Saklad, *Air Qual., Atmos. Health*, 2009, **2**, 185–198.
- 80 S. R. Magari, J. Schwartz, P. L. Williams, R. Hauser, T. J. Smith and D. C. Christiani, *Epidemiology*, 2002, **13**, 305–310.
- 81 K. Kim, E.-Y. Park, K.-H. Lee, J.-D. Park, Y.-D. Kim and Y.-C. Hong, *Environ. Health Prev. Med.*, 2008, **14**, 60–66.
- 82 T. Z. Maung, J. E. Bishop, E. Holt, A. M. Turner and C. Pfrang, *Int. J. Environ. Res. Public Health*, 2022, **19**, 8752.

