

12-2016

Indoor Inhalation Intake Fractions of Fine Particulate Matter: Review of Influencing Factors

Natasha Hodas

Portland State University, nhodas@pdx.edu

Miranda Loh

Institute of Occupational Medicine

Hyeong-Moo Shin

University of California, Davis

Dingsheng Li

Technical University of Denmark

Deborah Bennett

University of California

See next page for additional authors

Let us know how access to this document benefits you.

Follow this and additional works at: https://pdxscholar.library.pdx.edu/esm_fac

 Part of the [Environmental Sciences Commons](#)

Citation Details

Hodas, N., Loh, M., Shin, H., Li, D., Bennett, D., McKone, T. E., & ... Fantke, P. (2016). Indoor inhalation intake fractions of fine particulate matter: review of influencing factors. *Indoor Air*, 26(6), 836-856. doi:10.1111/ina.12268

This Article is brought to you for free and open access. It has been accepted for inclusion in Environmental Science and Management Faculty Publications and Presentations by an authorized administrator of PDXScholar. For more information, please contact pdxscholar@pdx.edu.

Authors

Natasha Hodas, Miranda Loh, Hyeong-Moo Shin, Dingsheng Li, Deborah Bennett, Thomas E. McKone, Olivier Jolliet, Charles J. Weschler, Matti J. Jantunen, Paul Lioy, and Peter Fantke

Review

Indoor inhalation intake fractions of fine particulate matter: review of influencing factors

Abstract Exposure to fine particulate matter (PM_{2.5}) is a major contributor to the global human disease burden. The indoor environment is of particular importance when considering the health effects associated with PM_{2.5} exposures because people spend the majority of their time indoors and PM_{2.5} exposures per unit mass emitted indoors are two to three orders of magnitude larger than exposures to outdoor emissions. Variability in indoor PM_{2.5} intake fraction ($iF_{in,total}$), which is defined as the integrated cumulative intake of PM_{2.5} per unit of emission, is driven by a combination of building-specific, human-specific, and pollutant-specific factors. Due to a limited availability of data characterizing these factors, however, indoor emissions and intake of PM_{2.5} are not commonly considered when evaluating the environmental performance of product life cycles. With the aim of addressing this barrier, a literature review was conducted and data characterizing factors influencing $iF_{in,total}$ were compiled. In addition to providing data for the calculation of $iF_{in,total}$ in various indoor environments and for a range of geographic regions, this paper discusses remaining limitations to the incorporation of PM_{2.5}-derived health impacts into life cycle assessments and makes recommendations regarding future research.

**N. Hodas^{1,2}, M. Loh³, H.-M. Shin⁴,
D. Li⁵, D. Bennett⁴, T. E. McKone^{6,7},
O. Jolliet⁵, C. J. Weschler^{8,9},
M. Jantunen¹⁰, P. Lioy^{8,*},
P. Fantke¹¹**

¹Division of Chemical Engineering, California Institute of Technology, Pasadena, CA, USA, ²Department of Environmental Science and Management, Portland State University, Portland, OR, USA, ³Institute of Occupational Medicine, Edinburgh, UK, ⁴Department of Public Health Sciences, University of California, Davis, CA, USA, ⁵Department of Environmental Health Sciences, University of Michigan, Ann Arbor, MI, USA, ⁶School of Public Health, University of California, Berkeley, CA, USA, ⁷Lawrence Berkeley National Laboratory, Berkeley, CA, USA, ⁸Environmental and Occupational Health Sciences Institute, Rutgers University, Piscataway, NJ, USA, ⁹International Centre for Indoor Environment and Energy, Technical University of Denmark, Kgs. Lyngby, Denmark, ¹⁰Department of Environmental Health, National Institute for Health and Welfare, Helsinki, Finland, ¹¹Department of Management Engineering, Technical University of Denmark, Kgs. Lyngby, Denmark

Key words: Fine particulate matter; Human exposure; Indoor air; Intake fraction; Life cycle impact assessment; Ventilation.

P. Fantke
Quantitative Sustainability Assessment Division
Department of Management Engineering
Technical University of Denmark
Produktionstorvet 424
2800 Kgs. Lyngby
Denmark
Tel.: +45 45254452
Fax: +45 45933435
e-mail: pefan@dtu.dk
*Deceased.

Received for review 10 July 2015. Accepted for publication 2 November 2015.

Practical Implications

This paper reviews and summarizes the factors that influence indoor inhalation intake fraction of fine particulate matter, with a focus on primary particle emissions indoors. It provides valuable data for the calculation of indoor inhalation intake fraction for a range of indoor environments and contributes to the effort to incorporate PM_{2.5}-derived health impacts into life cycle assessment.

Introduction

Human exposure to fine particulate matter (PM_{2.5}) is a major contributor to disease burden on a global scale (WHO, 2002, 2013). The indoor environment is a particularly important venue for exposure to PM_{2.5} because people spend the majority of their time indoors (Klepeis et al., 2001; Phillips and Moya, 2014 and references therein). Further, due to the lesser degree of dilution, chemical transformation, and dispersion, as well as the higher density of occupants indoors, exposures per unit mass of PM_{2.5} emitted indoors are two to three orders of magnitude larger than exposures to emissions to the outdoor environment (Ilacqua et al., 2007; Klepeis and Nazaroff, 2006; Lai et al., 2000; Nazaroff, 2008; Smith, 1988). To fully assess the impacts associated with all emission sources of PM_{2.5} and to evaluate the life cycle environmental performance of products and systems (e.g., energy and transport systems, food products and production systems, and consumer products), there is a need for the incorporation of PM_{2.5} exposures and the associated health effects into Life Cycle Impact Assessments (LCIA), with a specific need for the consideration of the impacts related to indoor exposures to PM_{2.5} emitted or formed indoors.

Due to current limitations in data availability and modeling tools that systematically combine indoor and outdoor intakes from indoor and outdoor sources, as well as challenges in consistently linking indoor and outdoor intakes to exposure–response, indoor sources and related intake of PM_{2.5} are currently not considered in product-related assessments (Humbert et al., 2015). To integrate indoor sources into such assessment frameworks, there is a need for (1) the identification of factors contributing substantially to variability in PM_{2.5} exposure and an examination of the value of accounting for this variability when assessing PM_{2.5} health impacts, (2) the aggregation and evaluation of modeling tools and data available for assessing human exposure to PM_{2.5}, and (3) a thorough assessment of the availability of exposure–response functions (ERFs) and the appropriateness of ERF shape (e.g., linear, nonlinear, presence of a threshold) for a variety of health outcomes (Fantke et al., 2015). With the aim of addressing these barriers and the lack of a standardized methodology to estimate exposures and health effects, the United Nations Environment Programme (UNEP)-Society for Environmental Toxicology and Chemistry (SETAC) Life Cycle Initiative formed a task force to provide guidance for the assessment of PM_{2.5} exposures and associated health effects (Fantke et al., 2015; Jolliet et al., 2014). Under the framework of this task force and with input from an international team of experts, this paper constitutes a first step toward incorporating indoor PM_{2.5} exposures into LCIA by characterizing the factors that drive variability in the

inhalation intake fraction of PM_{2.5} derived from indoor sources.

Inhalation intake fraction (*iF*), which is defined as the ratio of mass of a pollutant inhaled by an exposed human population to the total mass associated with a given source (Bennett et al., 2002), provides a well-suited metric by which to consider PM_{2.5} impacts in the context of LCIA. As an exposure metric, *iF* integrates components that are key to such assessments: (1) it describes source–receptor relationships in a manner that allows for direct comparisons across emission sources and (2) it can readily be related to potential toxicity in terms of specific health outcomes when exposure–response relationships are known (Bennett et al., 2002; Fantke et al., 2015; Ilacqua et al., 2007; Nazaroff, 2008). Table 1 illustrates the contributions of PM_{2.5} derived from indoor sources (*S*_{in}) and outdoor sources (*S*_{out}) to indoor intake, outdoor intake, total intake, and the intake fraction of PM_{2.5}. As is described in detail below, this paper reviews the major factors influencing the inhalation intake fraction of PM_{2.5} derived from indoor sources (Table 1, Equation 1). Examples of common indoor sources of PM_{2.5} include cooking, household and office appliances, smoking, cleaning, candles, and heating appliances or stoves. Additional efforts are currently underway within the UNEP-SETAC LCIA framework to characterize the other aspects of PM_{2.5} intake and intake fraction shown in Table 1.

Indoor inhalation intake fraction (*iF*_{in,total}) describes the total inhalation intake of PM_{2.5} (in kg) per unit mass emitted indoors (in kg). Two components contribute to *iF*_{in,total} (Table 1, Equation 1): (1) the fraction of PM_{2.5} emitted or formed indoors that is taken in via inhalation indoors (*iF*_{in→in}) and (2) the fraction of PM_{2.5} emitted or formed indoors that is transported outdoors and taken in via inhalation outdoors (*iF*_{in→out}). However, because PM_{2.5} of indoor origin experiences a greater degree of dispersion and dilution following transport outdoors and outdoor population density is lower than indoors, *iF*_{in→out} is typically three orders of magnitude smaller than *iF*_{in→in} (Humbert et al., 2011; Ilacqua et al., 2007; Klepeis and Nazaroff, 2006; Lai et al., 2000; Nazaroff, 2008; Smith, 1988). Thus, in calculations of *iF*_{in,total}, *iF*_{in→out} can be considered negligible compared to *iF*_{in→in}. As a result, this paper focuses on characterizing the major factors contributing to variability in *iF*_{in→in}, as this term dominates *iF*_{in,total}. While not the main focus, we also note the importance of interactions between pollutants of outdoor and indoor origin and the influence of outdoor PM_{2.5} sources on cumulative indoor intake (Table 1, Equation 2) and briefly discuss the current state of knowledge regarding these aspects.

Nazaroff (2008) divided the factors influencing variability in *iF*_{in→in} for primary particles into three categories: (1) factors related to building characteristics (e.g., ventilation, airflow, and mixing rates), (2) factors related to occupant characteristics and behaviors (e.g.,

Table 1 Matrix illustrating the contributions of PM_{2.5} derived from indoor and outdoor sources to indoor intake, outdoor intake, total intake, and intake fraction of PM_{2.5}

Emission (kg _{emitted} /h)	Indoor intake (kg _{makeup} /h)	Outdoor intake (kg _{makeup} /h)	Total intake (kg _{makeup} /h)	Intake fraction (kg _{makeup} /h per kg _{emitted} /h)
Indoor (PM _{2.5} or precursor) emissions S_{in}	Indoor intake due to indoor emissions $\underbrace{iF_{in \rightarrow in}}_{\left(\frac{IR_{in} \times \rho_{in}}{V_{in} \times K_{in}}\right)} \times S_{in}$	Outdoor intake due to indoor emissions $\underbrace{iF_{out \rightarrow out}}_{\left(\frac{IR_{out, total} \times f_{in \rightarrow out}}{V_{out} \times K_{out}}\right)} \times S_{in}$	Total intake due to indoor emissions $iF_{in, total} \times S_{in}$	Total intake due to indoor emissions per unit of indoor emissions $iF_{in, total} = \underbrace{\left(\frac{IR_{in} \times \rho_{in}}{V_{in} \times K_{in}}\right)}_{iF_{in \rightarrow in}} + \underbrace{\left(\frac{IR_{out, total} \times f_{in \rightarrow out}}{V_{out} \times K_{out}}\right)}_{iF_{out \rightarrow in}} \quad (1)$
Outdoor (PM _{2.5} or precursor) emissions S_{out}	Indoor intake due to outdoor emissions $\underbrace{iF_{in \rightarrow in}}_{\left(\frac{IR_{in} \times \rho_{in}}{V_{in} \times K_{in}}\right)} \times S_{out}$	Outdoor intake due to outdoor emissions $\underbrace{iF_{out \rightarrow out}}_{\left(\frac{IR_{out} \times \rho_{out}}{V_{out} \times K_{out}}\right)} \times S_{out}$	Total intake due to outdoor emissions $iF_{out, total} \times S_{out}$	
	Cumulative indoor intake due to indoor and outdoor emissions $\underbrace{iF_{in \rightarrow in}}_{\left(\frac{IR_{in} \times \rho_{in}}{V_{in} \times K_{in}}\right)} \times S_{in} + \underbrace{\left(iF_{in, total} \times f_{out \rightarrow in}\right)}_{\left(\frac{IR_{in} \times \rho_{in}}{V_{in} \times K_{in}}\right) \times f_{out \rightarrow in}} \times S_{out} \quad (2)$	Cumulative outdoor intake due to indoor and outdoor emissions $iF_{out \rightarrow out} \times S_{out} + iF_{in \rightarrow out} \times S_{in}$	Cumulative intake due to indoor and outdoor emissions $iF_{in, total} \times S_{in} + iF_{out, total} \times S_{out}$	

Aspects discussed in this paper are highlighted in grey and specific areas of focus are in red.

S_{in} or S_{out} , indoor or outdoor PM_{2.5} source emission rate; $iF_{in \rightarrow in}$, fraction of PM_{2.5} emitted/formed indoors that is taken in via inhalation indoors; $iF_{in \rightarrow in}$, fraction of PM_{2.5} emitted/formed indoors that is transported outdoors and taken in via inhalation outdoors; $iF_{out \rightarrow in}$, fraction of PM_{2.5} emitted/formed outdoors that is taken in via inhalation indoors; IR_{in} or IR_{out} , individual inhalation rate indoors or outdoors (m³ inhaled/h); ρ_{in} or ρ_{out} , number of exposed persons in an indoor or outdoor location; V_{in} or V_{out} , volume of indoor or outdoor location (m³); K_{in} or K_{out} , total indoor or outdoor particle removal rate attributable to all loss mechanisms (e.g., air exchange, particle deposition) (per h); $iF_{in, total}$, total indoor inhalation intake fraction; $iF_{out, total}$, total outdoor inhalation intake fraction; $f_{in \rightarrow out}$, fraction of indoor-generated (emitted/formed) PM_{2.5} transported outdoors; $f_{out \rightarrow in}$, fraction of outdoor-generated (emitted/formed) PM_{2.5} transported indoors. Note that there is no cumulative intake fraction.

inhalation rates and occupancy/activity patterns), and (3) pollutant dynamics (e.g., first order removal processes and sorptive interactions). That study noted the need for a ‘richly constituted tool kit to effectively comprehend the system of the human health risk associated with products and processes in indoor environments’. Humbert et al. (2011) provided an initial set of parameters characterizing two archetypal indoor environments [residences within the United States (USA) and mechanically ventilated offices]. Herein, we expand on that effort by developing an inventory of parameters (i.e., a ‘tool kit’) to (1) address each of the factors influencing $iF_{in \rightarrow in}$ discussed by Nazaroff (2008) and (2) allow for the characterization of multiple archetypal indoor environments (e.g., residences, offices, schools, etc.), covering a broad range of geographic scales.

Methods

For each category of factors influencing $iF_{in \rightarrow in}$ (building, occupant, and pollutant factors), subgroups with expertise in that specific field were created within an indoor air task force. Literature searches conducted by each subgroup were obtained from Web of Science, Google Scholar, and/or SCOPUS with search terms representing sources of variability related to the above-described categories (e.g., ‘air exchange rate measurements’, ‘building ventilation’, ‘commercial building ventilation rates’, ‘inhalation rates’, ‘indoor particle deposition’, ‘indoor particle emission rates’, etc.). When available, review papers were preferentially selected to be included in this review due to its multidimensional focus. Collected references were then reviewed and compiled to provide an inventory of data sources (e.g., peer-reviewed scientific articles and reports) and data regarding each factor influencing $iF_{in \rightarrow in}$. We included key papers (i.e., those with the soundest experimental/modeling practices, those that provide the greatest breadth of data, and those that allow for consideration of a range of exposure scenarios) in the present review and provide data from those papers in the Supporting Information. In general, the data compiled include summary statistics (i.e., mean, standard deviation, geometric mean, geometric standard deviation, percentiles, minimum, and maximum values) from individual studies conducted under a variety of experimental conditions and for a range of geographic locations. Where possible, data are categorized by country/geographic region and specific conditions to allow for the selection of data most relevant to an exposure scenario of interest. Each factor contributing to variability in $iF_{in \rightarrow in}$ is discussed in an individual section below.

Building factors

Building-specific factors influencing $iF_{in \rightarrow in}$ include building volume and ventilation (Table 1, Equations 1

and 2). Building ventilation is a key parameter in estimating $iF_{in \rightarrow in}$, as it drives the transport, dispersion, and dilution of PM_{2.5} emitted indoors. Indoor ventilation is driven by three processes: (1) leakage through cracks in the building shell and walls (infiltration/exfiltration), (2) airflow through open windows and doors (natural ventilation), and (3) mechanical ventilation (i.e., flow driven by fans; Chan et al., 2005; US EPA, 2011). Infiltration/exfiltration and natural ventilation are driven by pressure gradients that exist across the building envelope due to indoor–outdoor temperature differences and wind (US EPA, 2011). Mechanical ventilation systems range between exhaust- or supply-only systems (e.g., bathroom and kitchen exhaust fans/hoods), balanced supply and exhaust systems, localized unitary/single-zone systems, and central/integrated systems (Brelvi and Seppänen, 2011; Litiu, 2012; Sippola and Nazaroff, 2002). Building ventilation is typically quantified as whole-building/whole-zone air exchange rates (AERs) (per h) or, as is common for nonresidential/commercial buildings, volumetric flow rate normalized by building occupancy, volume, or floor area (l/s per person, l/s/m³, l/s/m²) (Persily, 2015). In the following paragraphs, we review the body of literature focused on characterizing these building properties and processes in a range of building archetypes.

Residential buildings. Residential ventilation rates have been most heavily studied in Europe (Asikainen et al., 2013; Dimitroulopoulou, 2012 and references therein; Hänninen et al., 2011; Orru et al., 2014) and North America (Figure 1a) (Bari et al., 2014; Breen et al., 2014; Chen et al., 2012; Clark et al., 2010; El Orch et al., 2014; MacNeill et al., 2012, 2014; Persily, 2015; Persily et al., 2010; US EPA, 2011 and references therein). While more limited in their number and scope, some studies have also been carried out in New Zealand (McNeil et al., 2012), Asia (Baek et al., 1997; Huang et al., 2014; Li and Li, 2015; Park et al., 2014; Shi et al., 2015; Williams and Unice, 2013), Africa, and South America (Williams and Unice, 2013 and references therein) (Figure 1a). In addition to those studying the housing stock in broad geographic regions, some studies have focused on homes with specific characteristics (e.g., new homes, energy-efficient homes, low-income/public housing; US EPA, 2011; Zota et al., 2005). A limited number of studies have characterized ventilation in homes in developing countries (L'Orange et al., 2015 and references therein; Williams and Unice, 2013) (Figure 1a). The use of solid fuels for cooking and heating, particularly in developing countries, is a leading indoor air quality issue on a global scale, with approximately 4.3 million premature deaths annually attributed to related pollutant exposures (www.WHO.int/indoorair/en). As a result, such measurements for homes in developing countries are very important to

the effort to incorporate the impacts of indoor PM_{2.5} exposures into LCIA.

The above-described body of work illustrates that there is spatial variability in residential ventilation with climate, building construction characteristics, home age, heating, ventilation, and air conditioning (HVAC) system configurations, ventilation standards and regulations, and residence type (i.e., detached, single-family homes, apartments) (Figure 2a). Temporal heterogeneity in ventilation rates results from variability in meteorological conditions and human behaviors such as window opening and mechanical ventilation system usage. The compilation of data characterizing homes over a broad range of geographic scales, housing types, seasons, and meteorological conditions is needed because the prevalence of different ventilation systems varies strongly across these factors. For example, AERs in 100% of both apartments and detached homes in Bulgaria are driven by infiltration and natural ventilation. On the other hand, 48% of detached homes in Finland have mechanical ventilation systems. This proportion increases to 72% when considering apartments (Litiu, 2012). To aid in the selection of representative ventilation parameters when calculating $iF_{in \rightarrow in}$, the ventilation rates and air exchange rate data provided here are categorized by country, home type, season, and ventilation system where the available data allow for this (Figures 1a and Supporting Information). Studies characterizing window-opening behavior and/or mechanical ventilation system usage and runtime (e.g., Breen et al., 2014; Chao, 2001; El Orch et al., 2014; Fabi et al., 2012; Gózeński et al., 2014; Iwashita and Akasaka, 1997; Johnson and Long, 2005; Levie et al., 2014; Marr et al., 2012; Persily, 2015; Stephens, 2015; Wallace et al., 2002; US EPA, 2011) provide needed information for accounting for temporal and spatial variability in ventilation conditions.

Figure 2a summarizes available residential air exchange rate data, with detailed data provided in the Supporting Information. For all residential AER measurements combined, we observed a median value of 0.50 per h [95% confidence interval (CI) = 0.08, 8.2 per h (Figure 2a), which is slightly higher than the recommended median value of 0.45 per h for homes in the USA provided in the Environmental Protection Agency Exposure Factors Handbook (US EPA EFH) (US EPA, 2011). This difference can likely be attributed, at least in part, to our inclusion of a small number of measurements from high AER homes in developing countries, as well as differences in home characteristics and ventilation systems across nations. While treated as a single distribution above for the purpose of comparison against the recommended value in the US EPA EFH, residential AERs are likely best characterized by a bimodal distribution. This is evidenced by differences in the median AER values for homes in developed and

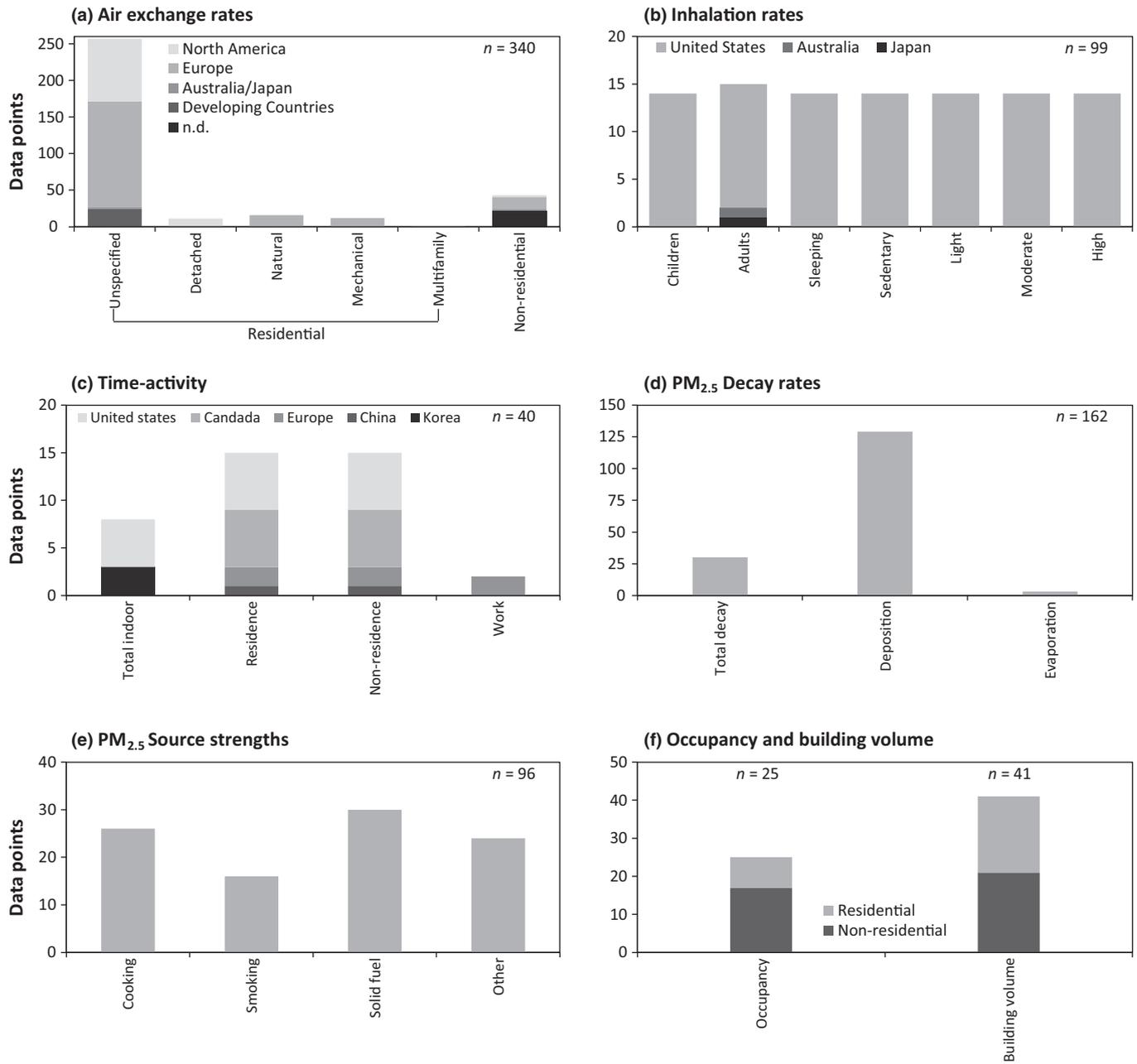


Fig. 1 Frequency plot illustrating the number of data points (i.e., measured or modeled value or summary statistic from a distribution of measurements describing the parameter of interest) gathered from the literature for the primary factors influencing indoor inhalation intake fraction of PM_{2.5}: (a) air exchange rates, (b) inhalation rates, (c) time-activity factors, (d) particle decay rates, (e) indoor PM_{2.5} source strengths, and (f) occupancy and building volume. (a) Air exchange rates are shown for detached/single-family homes ('Detached'), multifamily homes ('Multifamily'), homes without mechanical ventilation (i.e., infiltration and natural ventilation) ('Non-mechanical'), mechanically ventilated homes ('Mechanical'), homes in developing countries ('Developing'), residential buildings for which the above-described characteristics have not been specified ('Unspecified'), and non-residential buildings ('Non-residential'). (b) Inhalation rates are for adults, children, and by activity level (sleeping, sedentary, light, moderate, and high). (c) Time-activity factors include total hours spent indoors ('Total Indoors'), in the residence ('Residence'), in other indoor locations ('Non-residence'), and at work ('Work') per day. (d) Particle decay rates are for all particle loss mechanisms combined ('Total Decay') and for losses driven only by deposition. (e) Indoor PM_{2.5} emission source strengths include cooking, smoking, solid fuel combustion, and other indoor sources. (f) Occupancy and building volume data are categorized by residential and non-residential indoor environments. Where possible, data are categorized by country/geographic region (Not determined ('n.d.') means that geographic region is unspecified). Studies included here have primarily been conducted in North America and Europe (a, b, c). In addition, there are disparities in the types of indoor environments studied in previous work, with the majority of studies focusing on residential environments and a smaller number of studies considering industrial and commercial buildings

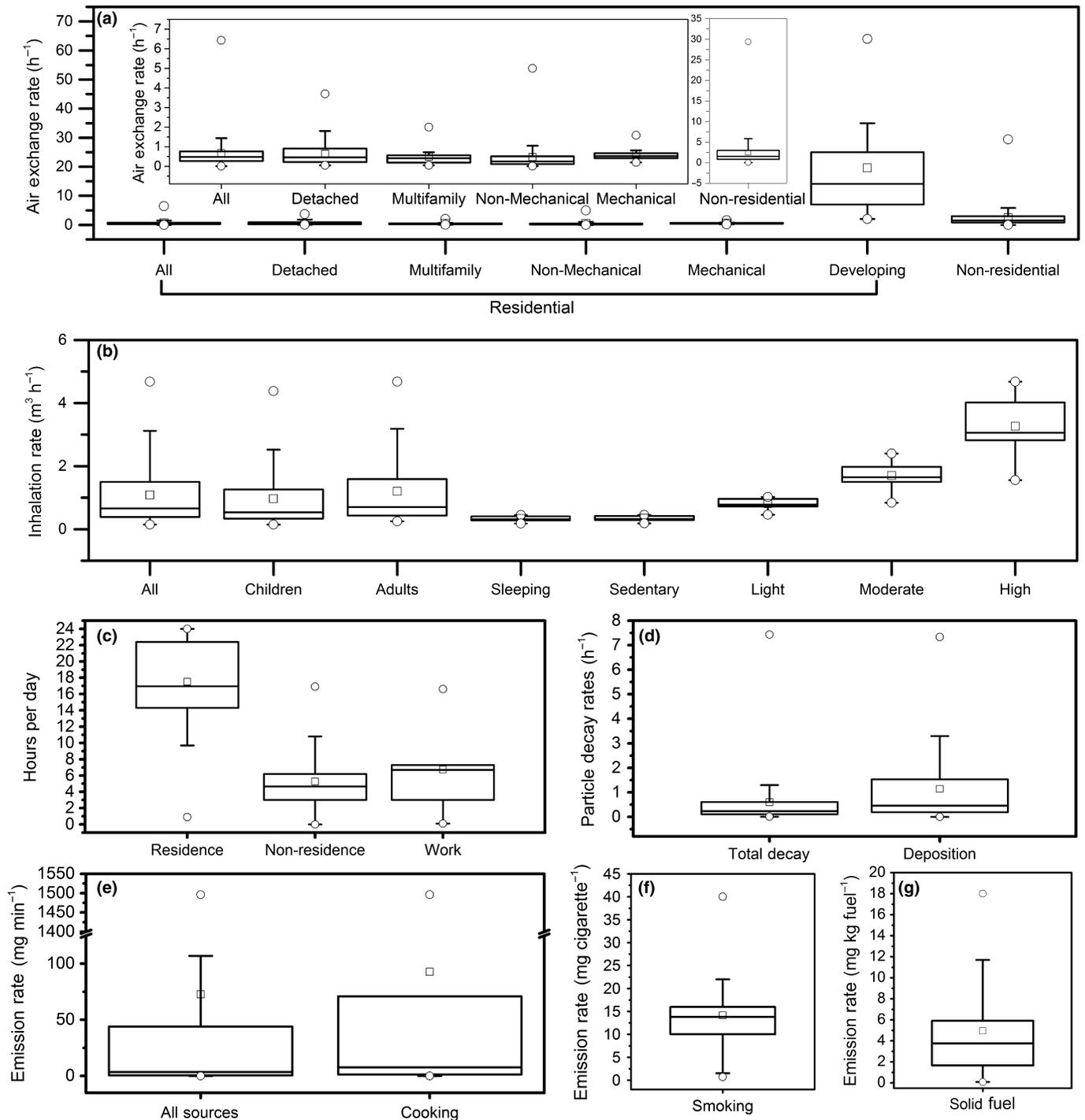


Fig. 2 Summary of measured or modeled values describing the parameter of interest for (a) building air exchange rates, (b) inhalation rates, (c) time-activity factors, (d) particle decay rates, and (e–g) indoor PM_{2.5} source strengths reported in the literature. For all plots, the boxes indicate the 25th percentile, median, and 75th percentile. Minimum and maximum values are indicated with circles and mean values are indicated with squares. (a) Air exchange rates shown are for all homes combined (excluding homes in developing nations) ('All') and separately for detached/single-family homes ('Detached'), multifamily homes ('Multifamily'), homes without mechanical ventilation (i.e., infiltration and natural ventilation) ('Non-mechanical'), mechanically ventilated homes ('Mechanical'), homes in developing countries ('Developing'), and non-residential buildings ('Non-residential'). (b) Inhalation rates are for all measurements combined ('All'), and separately for adults (>21 years), children (≤21 years), and activity level (sleeping, sedentary, light, moderate, and high). (c) Time-activity factors include hours per day spent in the residence ('Residence'), in other indoor locations ('Non-residence'), and at work ('Work'). (d) Particle decay rates are given for all particle loss mechanisms combined ('Total Decay') and for losses driven only by deposition. (e) Source emissions are given for common indoor PM_{2.5} sources including cooking, cleaning, smoking, and various appliances combined, excluding the combustion of solid fuels ('All Sources'). (e), (f), and (g) Source emissions are also illustrated for cooking, smoking, and solid fuel combustion separately. The total number of observations for each parameter is shown in Figure 1 and all underlying data are provided in the Supporting Information

developing countries: median (95% CI) = 0.48 (0.08–2.26) per h and 14.1 (2.0, 61.0) per h, respectively.

Many of the studies described above in which air exchange and ventilation are measured also provide data regarding the volume/floor area of the homes studied (Figure 1f). It is important to note that homes included are not necessarily statistically representative of the housing stock and this influences estimates of both home volume and ventilation. Population-level data describing home characteristics can also typically be gathered from census and housing survey databases (e.g., the American Census, American Housing Survey, Eurostat, and Census India). Recommended values for various housing and building characteristics are also available in reports summarizing exposure factors in several countries (Phillips and Moya, 2014 and references therein; US EPA, 2011). Available measurements of residential volumes illustrate their high variability, both within and across nations, with values ranging from 15–1446 m³ [median (95% CI) = 247 (41, 971) m³] (see Supporting Information). The median residential volume for the studies considered in this work is lower than the recommended value provided in the US EPA EFH (492 m³) (US EPA, 2011), likely illustrating differences in residential volumes across regions of the world.

Non-residential buildings. Ventilation measurements have been conducted in a range of non-residential buildings, including retail stores (Dutton et al., 2015; US EPA, 2011; Zaatari et al., 2014 and references therein), schools, kindergartens, and daycare centers (Aelenei et al., 2013; Brelih and Seppänen, 2011; Canha et al., 2013; Coley and Beisteiner, 2002; Emmerich and Crum, 2006; Guo et al., 2008; Li et al., 2007; Mi et al., 2006; Santamouris et al., 2008; Sundell et al., 2011; Wargocki et al., 2002) offices (Dimitroulopoulou and Bartzis, 2013; Persily and Gorfain, 2004), fitness facilities (Zaatari et al., 2014), jails (Li et al., 2007; Seppänen et al., 1999), and healthcare facilities, hospitals, and nursing homes (Li et al., 2007 and references therein; Wargocki et al., 2002). Summary statistics of more than 700 measurements from 17 studies, for example, have been compiled for retail facilities, bars/restaurants, healthcare facilities, fitness facilities, offices, and schools (Zaatari et al., 2014). As is true for residential ventilation rates, measurements in non-residential buildings are more heavily focused in North America and Europe, with a smaller number of studies also conducted in Asia (Figure 1a). Non-residential AERs are summarized in Figure 2a, with more detailed information (e.g., categorized by building type) provided in the Supporting Information. We observed a median AER for non-residential buildings of 1.5 per h (95% CI = 0.29, 9.1 per h).

The above-described studies again demonstrate geographic variability in ventilation system characteristics

and the prevalence of mechanically and naturally ventilated buildings, as well as temporal variability in ventilation with meteorological conditions, window opening, and HVAC system operation. For example, 100% of schools and kindergartens are naturally ventilated in Italy, while only 5% and 28% of kindergartens and schools are naturally ventilated in Finland (Litju, 2012). Sippola and Nazaroff (2002) note that single-zone HVAC systems are common in smaller commercial buildings with floor areas on the order of 150 m², while central systems dominate in larger buildings (>1000 m²) such as malls, university buildings, theaters, and retail centers.

A small number of studies discuss window-opening and HVAC system-use behavior in commercial/non-residential buildings (e.g., D'Oca and Hong, 2014; Fabi et al., 2012; Li et al., 2015; Ramos and Stephens, 2014; Roetzel et al., 2010; Stephens, 2015). Two recent studies (Bennett et al., 2012; Chan et al., 2014) conducted detailed measurements of AERs and ventilation rates in thirty-seven commercial buildings and nineteen retail stores, respectively, and provided summary statistics for various building types (e.g., grocery stores, hardware stores, restaurants, healthcare facilities, and public assembly spaces) and for varying ventilation conditions (e.g., with doors open/closed, with and without mechanical ventilation systems in use).

As was true for the residential ventilation studies, many of the above-described studies provide information regarding the characteristics of the buildings studied, including building volume and/or floor area; however, again, these values are typically not statistically representative of the full range of non-residential building stock. The Building Assessment Survey and Evaluation (BASE) Study provides measurements of building and occupied-space size for 100 randomly selected large office buildings in the USA (Persily and Gorfain, 2004). US EPA (2011) is also a valuable resource for summary statistics of volume data for buildings with a wide range of uses and sizes (e.g., warehouses, shopping malls, schools, and healthcare facilities). As a result of the range of building uses, commercial building volumes display a large degree of variability, ranging from 408 to 849 505 m³ [median (95% CI) = 3398 (461, 192 554) m³] (see Supporting Information).

Inter- and intrazonal airflows and mixing. Interzonal and intrazonal airflow and local-scale mixing (i.e., convective and advective mixing on intrazonal scales) can be of importance in both residential and non-residential indoor environments, specifically when considering differences in exposures and $iF_{in \rightarrow in}$ for building occupants with varying proximities to sources of interest (Drescher et al., 1995; Nazaroff, 2008). Measurements of interzonal and intrazonal flows are limited. In addition, these flows vary within and across buildings and

depend on multiple factors including door opening, ventilation conditions, home layout, and temperature gradients (Klepeis, 2004; McGrath et al., 2014). Thus, selecting a representative value or sampling from a distribution of measured values when calculating $iF_{in \rightarrow in}$ is not straightforward. As a result, such flows typically must be modeled for an exposure scenario of interest.

Commonly used models for the estimation of interzonal flows include COMIS (Feustel, 1998) and CONTAM (Walton and Dols, 2013). AER and interzonal flows predicted with CONTAM and/or COMIS have been evaluated against measurements conducted in more than ten countries and for a variety of building types (Emmerich, 2001 and references therein; Emmerich et al., 2004; Haas et al., 2002). Details regarding the required inputs and use of these models are available in their respective users' manuals (Feustel, 1998; Walton and Dols, 2013).

Computational fluid dynamics (CFD) has been used to explicitly model airflow and turbulence on smaller, within-room scales (e.g., Gadgil et al., 2003; Zhang and Chen, 2007; Zhao and Guan, 2007; Zhao et al., 2008). Pragmatically, multizone and zonal modeling methods can be combined by nesting an intrazonal model within an interzonal model (Stewart and Ren, 2003, 2006; Wang and Chen, 2007), so that a specific room of interest (e.g., the room with a PM_{2.5} source) can be divided into several small zones, while other rooms within the same home/building are treated as larger, well-mixed zones.

Alternatively, Bennett and Furtaw (2004) provide an estimate of a room-to-room air exchange rate distribution (mean = 3 per h, coefficient of variation = 0.30) based on measurements conducted under varying ventilation conditions within a single house. Du et al. (2012) characterized overall and season-specific interzonal airflows between living areas and bedrooms in 126 homes in Detroit, MI as the percentage of room-specific air exchange attributable to air entering from another zone. Along the same lines, Hellweg et al. (2009) suggest ranges of values for within-zone mixing factors (0.1 to 1.0) and interzonal air exchange rates (3 to 30 m³/min). These are examples of midway approaches between the typical single, well-mixed compartment assumption and more complex approaches based on CFD. Understanding the influence of smaller scale flows on $iF_{in \rightarrow in}$ is an important area of future research, with a rate coefficient representing the airflow between zones (including the near-person zone and the rest of an indoor environment) being a resulting metric of interest for use in LCIA.

Human exposure factors

Inhalation rate. Inhalation intake fraction is directly related to the inhalation rate (*IR*) of the subjects or population of interest (Table 1, Equation 1). Inhalation rates vary within and across individuals with

multiple factors including age, sex, body weight, and fitness and activity levels (Figure 2b) (US EPA, 2011). Studies quantifying *IR* are largely based on relationships between oxygen uptake and consumption, metabolism, and energy expenditure (US EPA, 2011). Using various methods to quantitate energy expenditure and oxygen consumption, multiple studies have measured *IR* for broad, representative populations (e.g., Jang et al., 2014a; Richardson and Stantec, 2013; US EPA, 2011 and references therein), while others have focused on specific populations of interest (US EPA, 2011 and references therein). Recommended values of *IR* for the general population categorized by age, gender, and activity level are available for the USA (US EPA, 2011), Canada (Richardson and Stantec Consulting Ltd, 2013), and Korea (Figure 1b) (Jang et al., 2014a). As is discussed below, materials are available to allow for the estimation of *IR* for populations for which such measurements have not been conducted. Specific populations of interest for which *IR* studies have been conducted include children, adults and children with asthma, and pregnant and lactating adult and adolescent women (US EPA, 2011). Such studies allow for the consideration of $iF_{in \rightarrow in}$ for susceptible populations or during specific periods of susceptibility.

Inhalation rates are commonly reported as long-term (m³/day), or short-term (m³/min) rates. The latter allow for distinguishing differences in *IR* arising from different levels of activity. When assessing chronic exposures, long-term *IR*s can be utilized to characterize $iF_{in \rightarrow in}$; however, short-term *IR*s are needed when considering acute exposures or exposures associated with a particular activity (i.e., where the emission is represented by a pulse rather than a continuous term). Short-term *IR*s are generally categorized by age, sex, and intensity of activity (e.g., resting/napping, sedentary, and light, moderate, and high intensity; Adams, 1993; US EPA, 2011). Some studies are as specific as to provide activity-level-specific, short-term *IR*s for activities conducted in the indoor environment (US EPA, 2011).

To use short-term *IR*s in estimates of $iF_{in \rightarrow in}$, information regarding the fraction of time spent at various activity levels is needed. As is discussed in more detail below, time-activity patterns have been documented for populations from a wide range of geographic regions [e.g., Australian Centre for Human Health Risk Assessment, 2012; Jang et al., 2014b; Klepeis et al., 2001; Statistics Canada, 2011; ExpoFacts (<http://expofacts.jrc.ec.europa.eu/>)] (Figure 1b). US EPA (2011) also provides age-specific estimates of time spent at various levels of activity intensity. The populations for which short-term *IR*s have been quantified are limited (Jang et al., 2014b; US EPA, 2011). Time-activity datasets can be combined with available short-term *IR* to predict *IR* distributions for populations for which such measurements are not available;

however, it must be acknowledged that there is greater uncertainty in these values. Sensitivity analyses may be valuable for evaluating the influence of this uncertainty in $iF_{in \rightarrow in}$. Several exposure factor reports detail population demographics and physiological conditions, which can then be used to generate population-specific long- and short-term *IR* distributions from available measurements (Phillips and Moya, 2014 and references therein). Figure 2b summarizes the results of key *IR* studies, with detailed data provided in the Supporting Information. Overall, average *IRs* for children, adults, and all age groups for the data gathered here are slightly higher than that provided in the US EPA EFH (0.97, 1.20, and 1.09 m³/h vs. 0.81, 1.04, and 0.92 m³/h). Median values (and 95% CI) of the data provided herein for *IRs* for children, adults, and all age groups are 0.55 (0.17, 3.40), 0.70 (0.26, 4.47), and 0.66 (0.22, 4.23) m³/h, respectively.

Time-activity patterns. In addition to serving as a predictor of activity intensity and *IR*, time-activity data provide valuable information regarding the time spent indoors and in various indoor locations. For a given subject, the cumulative intake of PM_{2.5} is a function of the time spent by that subject in various microenvironments (e.g., indoor locations) and the PM_{2.5} concentration profiles he or she is exposed to in each of those microenvironments. Thus, the characterization of activity patterns is crucial to estimating $iF_{in \rightarrow in}$. Studies characterizing time-activity patterns generally utilize diaries in which a representative sample of individuals from the general population record their activities over a 24 or 48 h period. The Center for Time Use Study at the University of Oxford provides a database of time-activity diary studies for approximately 100 countries in Africa, Asia, Australia, Europe, North America, and South America (Fisher and Tucker, 2013). Data from multiple nations are harmonized to allow for comparison across countries. In addition to references and links for the studies, where available, this database provides important information such as temporal scale of the study, sampling and data-collection methodology, sample size, and response rates. Some studies provide broader information that is useful for long-term exposure studies (e.g., total time spent indoors and time spent in the residence; Figures 1c and 2c), while others provide more detailed data, including time spent in various types of indoor environments (e.g., home, school, retail stores, etc.), time spent in different rooms within a residence, and time spent engaged in activities of relevance to specific PM_{2.5} emissions sources (e.g., cleaning, cooking; Jang et al., 2014b; Matz et al., 2014; Schweizer et al., 2007; US EPA, 2011; Zhao et al., 2009). Such studies have demonstrated that time-activity patterns vary with age, gender, location of residence (e.g., urban vs. rural), and various demographic and socioe-

conomic factors. Time-activity data are generally categorized by these factors and, thus, activity patterns can be estimated for a population of interest when demographic information is known. For the USA, the Consolidated Human Activity Database (CHAD; <http://www.epa.gov/heasd/chad.html>) brings together data from various studies, resulting in several thousand daily diaries that can be used in exposure simulation studies. The advantage of CHAD over other time-use databases is that it is developed specifically for exposure studies and certain parameters, such as time spent in indoor microenvironments, can be more easily distinguished. The Stochastic Human Exposure and Dose Simulation (SHEDS) Model (Burke et al., 2001), for example, simulates a population representative of the study populations, as well as their activity patterns, by sampling from input demographic data and CHAD.

Occupancy. Also key to determining $iF_{in \rightarrow in}$ is knowledge regarding the total number of people occupying a space influenced by indoor PM_{2.5} emissions (Nazaroff, 2008). Higher occupancy means a larger number of people in proximity to indoor sources and, thus, a higher population $iF_{in \rightarrow in}$. Several studies provide information regarding household size and composition, which can be utilized to estimate residential occupancy in calculations of $iF_{in \rightarrow in}$ (Figure 1f). The US Census Bureau (USCB), for example, provides information regarding the number and percentage of homes with household sizes ranging from one person to seven or more people, as well as demographic data describing households of varying sizes (USCB, 2010; Vespa et al., 2013). Similar information is available for the European Union (EU) and individual EU nations from Eurostat (2014). Bongaarts (2001) presented household size and composition for the developing countries based on surveys conducted in forty-three nations in the 1990s, but notes that household-size dynamics can change with increased urbanization and industrialization, trending toward smaller household sizes (i.e., trending toward the nuclear family). That study provided data regarding household size and the demographic characteristics of home occupants for four regions: Asia, Latin America, Near East/North Africa, and sub-Saharan Africa (see Supporting Information). Drivers of within- and between-nation/region variability are discussed and include level of development (e.g., gross national product) and residence in urban vs. rural areas. The United Nations Demographic Yearbook is a valuable reference for identifying and locating household occupancy and characteristic data collected through national censuses (United Nations, 2013). For non-residential buildings, US EPA (2011) provides distributions of employee numbers for commercial buildings with a wide range of uses (Supporting Information).

Pollutant-specific factors

Concentrations of PM_{2.5} and related intake in a given indoor environment or zone within an indoor environment depend on source emissions rates (S_{in}), as well as the removal mechanisms acting on the particles (k_{in}) (Table 1, Equation 2). Such removal mechanisms include the ventilation and transport processes discussed above, particle deposition, filtration in HVAC system filters and air cleaners, and, in some cases, chemical transformations/phase changes (Nazaroff, 2004). AERs and ventilation rates can be estimated using the data discussed above. In the following paragraphs, we discuss the data and tools available to take into account other factors influencing indoor PM_{2.5} concentrations and $iF_{in \rightarrow in}$, with a primary focus on PM_{2.5} emitted directly from indoor sources.

Indoor PM_{2.5} emissions. Multiple studies have characterized total PM_{2.5} emissions from common indoor sources and activities such as cooking, cleaning, smoking, use of various home and office appliances, candles, incense, and insect repellent coils (Figure 1e) (e.g., Afshari et al., 2005; Evans et al., 2008; Guo et al., 2004; He et al., 2004; ; He et al., 2007; Jetter et al., 2002; Lee and Wang, 2004; Liu et al., 2003; Lung and Hu, 2003; Olson and Burke, 2006; See and Balasubramanian, 2011; Torkmahalleh et al., 2012). Substantial variability in PM_{2.5} emission rates has been observed within and across sources (Figures 2e–g). For example, cooking activities can lead to emission rates as high as 467 mg/min (Olson and Burke, 2006), while emissions from printers were reported to be 2.8×10^{-4} mg/min (He et al., 2007). He et al. (2004) observed a median emission rate of 2.7 mg/min for frying food, while Olson and Burke (2006) reported a value of 6 mg/min. Emission rates for cooking activities vary with the cooking method (e.g., frying, grilling, baking), with the type of food or oils used in the cooking process (He et al., 2004; Olson and Burke, 2006; Torkmahalleh et al., 2012), and with stove type and the source of fuel (e.g., biomass, coal, gas, electric) (Supporting Information) (Jetter and Kariher, 2009; Jetter et al., 2012). The importance of a given source in terms of its contribution to $iF_{in \rightarrow in}$ varies with a variety of factors including the indoor environment under consideration, occupant activities, and time of day or season. For example, in office environments, appliances (e.g., printers, copy machines) may contribute substantially to indoor PM_{2.5} concentrations, while cooking, a major source in residential environments, is unlikely to be of importance. On the other hand, cleaning products are likely to be significant sources of PM_{2.5} in both office and residential environments.

The influence of specific PM_{2.5} sources on $iF_{in \rightarrow in}$ also varies geographically. Solid fuel combustion, for

example, is a particularly important source of indoor PM_{2.5} emissions in the developing world. As noted above, the effects of indoor exposures to solid fuel combustion emissions are a major global environmental health concern (www.who.int/indoorair/en). As a result, controlled laboratory studies and field measurements have been undertaken to characterize PM_{2.5} emissions from various cook stoves and fuel sources (Edwards et al., 2014 and references therein; Habib et al., 2008). It is important to note, however, that there is evidence that emissions rates measured in a laboratory setting differ from those in the field (Edwards et al., 2014) and future efforts are more focused on characterizing emissions in actual household settings. In addition to emissions, data regarding the percentage of households using solid fuels and geographic differences in fuel and stove use are available for estimating $iF_{in \rightarrow in}$ associated with solid fuel use (Bonjour et al., 2013; Rehfuess et al., 2006; www.who.int/indoorair/health_impacts/he_database/en; see Supporting Information).

As is discussed in more detail below, particle loss rates vary with particle size and, thus, information regarding the size distributions of particles emitted from specific sources is useful for calculating $iF_{in \rightarrow in}$. Recent work has provided particle size distributions and/or size-resolved emissions rates for a range of common indoor activities or sources including cooking (Abt et al., 2000; Hussein et al., 2006; Li and Hopke, 1993; Long et al., 2000; Ogulei et al., 2006; Wallace, 2006; Wallace et al., 2004), cleaning (Abt et al., 2000; Gehin et al., 2008; Kleeman et al., 1999; Long et al., 2000; Ogulei et al., 2006), candles, incense, and aroma lamps (Gehin et al., 2008; Hussein et al., 2006; Kleeman et al., 1999; Li and Hopke, 1993; Wallace, 2006), smoking (Hussein et al., 2006; Li and Hopke, 1993; Nazaroff, 2004), cook-stove use in developing countries and residential wood combustion (Armendáriz-Arnez et al., 2010; Hays et al., 2003; Kleeman et al., 1999; Shen et al., 2011), fuel-combustion lamps and appliances (Apple et al., 2010; Wallace, 2006), personal care products/appliances (e.g., hairspray, blow dryer) (Hussein et al., 2006), and printers (Gehin et al., 2008; Stephens et al., 2013; Wang et al., 2012).

Particle losses: deposition. Particle deposition describes all particle losses driven by Brownian diffusion, gravitational settling, interception, and impaction. Brownian diffusion dominates particle losses for particles with diameters smaller than about 0.1 μm [ultrafine particles (UFP)], while for larger particles, interception, impaction, and gravitational settling are the dominant loss processes (Finlayson-Pitts and Pitts, 2000). As a result, deposition loss rate coefficients [k_{dep} (per h)] vary with particle size (Hering et al., 2007; Long et al., 2001; Nazaroff, 2004; Ozkaynak et al., 1997; Riley et al., 2002). Multiple studies have measured par-

particle-size-resolved values of k_{dep} or indoor particle decay rates (i.e., the sum of all loss mechanisms) (e.g., Abt et al., 2000; Ferro et al., 2004; He et al., 2005; Howard-Reed et al., 2003; Long et al., 2001; Meng et al., 2007; Ozkaynak et al., 1997; Sarnat et al., 2006; Stephens and Siegel, 2013; Thatcher and Layton, 1995; Thatcher et al., 2003). These studies have been conducted under a range of sampling and building ventilation conditions. In addition to their particle size dependence, k_{dep} values vary with airflow conditions and indoor environment surface-to-volume ratios driven by the presence of furnishings and carpets (Howard-Reed et al., 2003; Lai, 2002; Nazaroff, 2004; Thatcher et al., 2002). For example, Thatcher et al. (2002) demonstrated that k_{dep} could vary by as much as a factor of 2.6 across different surface-to-volume (i.e., room-furnishing) scenarios and by as much as a factor of 2.4 with different values of airflow speed. Zhang et al. (2014) brings attention to the fact that variability in k_{dep} to surfaces with varying orientations (e.g., horizontal vs. vertical surfaces) can influence indoor $\text{PM}_{2.5}$ concentrations and $iF_{\text{in} \rightarrow \text{in}}$. That study provides vertical- and horizontal-surface deposition rates for particles in two broad $\text{PM}_{2.5}$ size classes.

Measurements conducted under various conditions have been combined and fit with a polynomial regression that describes k_{dep} as a function of particle size (Nazaroff, 2004; Riley et al., 2002). This fit does not take into account variability with ventilation conditions, room turbulence, surface-to-volume ratios, or room surface orientations; however, Hodas et al. (2014) found that indoor concentrations of ambient $\text{PM}_{2.5}$ modeled using k_{dep} values selected with this regression curve were well-correlated with measured indoor $\text{PM}_{2.5}$. El Orch et al. (2014) combined measurement data from multiple studies to predict particle-size-resolved k_{dep} values, fit a curve describing k_{dep} as a function of particle diameter, and developed a method to account for increased indoor airflow speeds when windows are open. In those circumstances, values of k_{dep} selected from curves describing depositional loss rates as a function of particle size (e.g., using Monte Carlo methods to sample from a particle size distribution) can be multiplied by 1.7 for windows open a large amount and by 1.23 when windows are open a small amount. In addition, a small number of studies have quantified deposition or decay rates for total $\text{PM}_{2.5}$ (Figures 1d and 2d) (He et al., 2005; Olson and Burke, 2006; Ozkaynak et al., 1997; Wallace et al., 2013). Such information can be useful in circumstances in which particle size distribution data are not available.

Particle losses: filtration. For homes with HVAC systems, particle losses will also be related to HVAC system recirculation rates and filter removal efficiencies. Several studies have measured size-resolved particle fil-

tration efficiencies for various filters commonly found in residential and commercial HVAC systems (Azimi et al., 2014; Hanley et al., 1994; Stephens and Siegel, 2012, 2013; Stephens et al., 2011). Stephens et al. (2011) also studied recirculation rates in residential and light-commercial HVAC systems. El Orch et al. (2014) extended this type of analysis to provide size-resolved filtration efficiencies for five classifications of filters, as well as estimates of the prevalence of these filter categories in homes. Waring and Siegel (2008) and Stephens and Siegel (2013) considered the influence of not only filtration, but also losses to heat exchangers and ducts within HVAC systems. Similarly, Sippola and Nazaroff (2002) reviewed studies of particle deposition in HVAC system ducts. Such losses are likely to be of particular importance in schools and commercial buildings. Filtration and fractional loss curves generated from such measurements have been used in many studies to estimate particle removal efficiencies as a function of particle size (Hodas et al., 2012, 2014; Riley et al., 2002).

HVAC system air recirculation rates are also key parameters in characterizing filtration rates. Recommended values for HVAC recirculation rates in residences (El Orch et al., 2014; Stephens, 2015; Stephens et al., 2011) and in non-residential buildings (Fadeyi et al., 2009; Sundell et al., 1994; Weschler et al., 1996; Zuraimi et al., 2007 and references therein) are available from a limited number of studies. Note also that the fraction of air that is recirculated in HVAC systems displays large spatial variability. Zuraimi et al. (2007), for example, state that 90% of air in conditioned office buildings in the USA and Singapore is recirculated. In some countries (e.g., Denmark and Germany), however, all mechanical ventilation systems must be single pass (i.e., no air is recirculated). Similarly, HVAC system runtimes directly govern whether or not a system is in operation and filtering particles at a given point in time, but like recirculation rates, measurements are limited (Stephens et al., 2011; Thornburg et al., 2004).

The prevalence of central air and heating systems is commonly documented in housing and energy surveys. US EPA (2011), for example, provides information regarding the prevalence of central heating and cooling systems in residential and commercial buildings. It is important to note, however, that the prevalence of central and recirculating HVAC systems is highly variable both within and across nations and geographic regions. The importance of collecting data regarding the heating and cooling systems (or lack thereof) present in households on a global scale has recently been highlighted (United Nations, 2008).

Particle resuspension. The resuspension of particles that have deposited on surfaces in indoor environments can also influence indoor $\text{PM}_{2.5}$ concentrations and $iF_{\text{in} \rightarrow \text{in}}$ (Ferro et al., 2004; Lioy, 2006 and references

therein). While typically considered to be an important determinant of exposures to particles larger than PM_{2.5}, Ferro et al. (2004) found that resuspension can result in the equivalent of a PM_{2.5} source strength ranging from 0.03 to 0.5 mg/min. The prevalence and magnitude of resuspension are dependent on the activities of building occupants, specifically cleaning (e.g., dusting, vacuuming) and active movement (e.g., walking, dancing, playing) (Ferro et al., 2004; Lioy, 2006). Thus, the influence of resuspension on $iF_{in \rightarrow in}$ is expected to vary temporally and spatially.

Transformation: phase changes and indoor chemistry. Phase changes and chemical transformation can lead to both increases and decreases in indoor PM_{2.5} concentrations. The partitioning of semivolatile organic compounds (SVOCs) between the gas and particle phases, for example, is dependent on indoor air temperature and the availability of particle-phase organic matter for sorption (Pankow, 1994). Thus, the extent to which a given indoor source of SVOCs contributes to $iF_{in \rightarrow in}$ will depend on the fraction of emissions from that source found in the particle phase, which, in turn, is dependent on the conditions of the indoor environment (i.e., temperature, organic PM_{2.5} concentrations). Examples of indoor sources of SVOCs that display this behavior include environmental tobacco smoke, flame retardants, plasticizers, and pesticides (Bennett and Furtaw, 2004; Gurunathan et al., 1998; Liang and Pankow, 1996; Lioy, 2006; Weschler and Nazaroff, 2008 and references therein). Estimating shifts in partitioning requires knowledge regarding volatility and partitioning coefficients of chemical species commonly found indoors, as well as the development of simplified models to predict SVOC partitioning in indoor air. This is an active area of research (Hodas and Turpin, 2014; Liu et al., 2014; Weschler, 2011; Weschler and Nazaroff, 2008, 2010); however, further work is needed to characterize semivolatile species of indoor origin before this process can be consistently incorporated into estimates of $iF_{in \rightarrow in}$.

The formation of secondary organic aerosols (SOA) from reactions between oxidants and gas-phase compounds emitted indoors can also substantially influence PM_{2.5} concentrations and $iF_{in \rightarrow in}$ (Long et al., 2000; Wainman et al., 2000; Waring, 2014; Waring and Siegel, 2010, 2013; Waring et al., 2011; Weschler, 2006, 2011; Weschler and Shields, 1999). Most work in this area has focused on reactions between terpenoids emitted from air fresheners, cleaning products, and scented personal care products and ozone (Nazaroff and Weschler, 2004; Singer et al., 2006; Waring and Siegel, 2010, 2013; Waring et al., 2011; Weschler, 2006, 2011). Such studies have demonstrated that indoor SOA formation varies with multiple factors including the chemicals present in indoor air, relative humidity, time of day, season, indoor ventilation conditions and HVAC

system use, indoor surface area and surface materials, and geographic location (Waring and Siegel, 2010; Weschler, 2011; Waring and Siegel, 2013; Youssefi and Waring, 2014). Indoor sources of ozone include photocopiers, laser printers, and electrostatic air cleaners; however, the majority of ozone present indoors is the result of transport from the outdoor environment (Weschler, 2000). SOA generated through reactions between VOCs of indoor origin and ozone of outdoor origin illustrates one mechanism through which interactions between indoor- and outdoor-generated pollutants can influence the intake of PM_{2.5} attributable, at least in part, to indoor sources. This complication of separating outdoor- and indoor-source contributions to the intake of PM_{2.5} in indoor environments is discussed further in the next section.

Influence of outdoor-generated pollutants on cumulative indoor intake of PM_{2.5}. The cumulative intake of PM_{2.5} that occurs indoors is influenced by both indoor and outdoor PM_{2.5} sources (Table 1, Equation 2) and depends on (1) primary emissions of PM_{2.5} from indoor sources, (2) the formation of secondary PM_{2.5} from precursors of indoor origin, (3) the transport of outdoor-generated PM_{2.5} into the indoor environment, and (4) interactions between pollutants of indoor and outdoor origin. This latter factor includes SOA formation through reactions of indoor-emitted volatile organic compounds (VOCs) and outdoor-generated oxidants, as well as the partitioning of outdoor-generated gas-phase SVOCs to particulate matter of indoor origin and/or the partitioning of gas-phase SVOCs emitted by indoor sources to outdoor-generated particles that have infiltrated indoors. Prior sections focused on factors (1) and (2). Below, we briefly explore the current state of knowledge regarding interactions between pollutants of outdoor and indoor origin and the influence of outdoor PM_{2.5} sources on cumulative indoor intake.

Outdoor-generated PM_{2.5} (ambient PM_{2.5}) that penetrates into and persists in the indoor environment is a major source of indoor PM_{2.5}. Multiple studies have quantified the fraction of ambient PM_{2.5} found in indoor air ($f_{out \rightarrow in}$) (Chen and Zhao, 2011 and references therein; Diapouli et al., 2013 and references therein). These studies have demonstrated that there is substantial between- and within-home variability in $f_{out \rightarrow in}$ (Allen et al., 2012; Hänninen et al., 2013; Kearny et al., 2014; MacNeill et al., 2012, 2014; Meng et al., 2005; Ott et al., 2000; Ozkaynak et al., 1997; Polidori et al., 2006; Weisel et al., 2005), illustrating the difficulty in utilizing measured values of $f_{out \rightarrow in}$ to estimate contributions of ambient PM_{2.5} to cumulative indoor intake. In addition, most studies are limited in their geographic and temporal scope and cannot be generalized to a broader population of homes. Two exceptions are the studies conducted by Hänninen

et al. (2011) and El Orch et al. (2014). Estimates of $f_{\text{out} \rightarrow \text{in}}$ for homes in ten European countries sampled as part of six studies were aggregated and summary statistics of $f_{\text{out} \rightarrow \text{in}}$ were provided for various climatic regions of Europe (Northern, Central, and Southern Europe) and by season (Hänninen et al., 2011). El Orch et al. (2014) conducted a detailed modeling study in which particle-size-resolved distributions of $f_{\text{out} \rightarrow \text{in}}$ for single-family homes in the USA were calculated.

For a given exposure scenario, $f_{\text{out} \rightarrow \text{in}}$ can also be calculated using a mass balance model in which indoor ambient $\text{PM}_{2.5}$ concentrations are described as function of AER, the efficiency with which particles penetrate across the building envelope, particle deposition, filtration in HVAC system filters and air cleaners, and, for semivolatile species, phase changes in indoor air (e.g., Hering et al., 2007; Hodas et al., 2012, 2014). Similarly, these physical and chemical processes also govern the outdoor transport of indoor-generated $\text{PM}_{2.5}$ and, thus, $iF_{\text{in} \rightarrow \text{out}}$ and $iF_{\text{in}, \text{total}}$ (see Table 1). While the contributions of $iF_{\text{in} \rightarrow \text{out}}$ to $iF_{\text{in}, \text{total}}$ are typically negligible compared to that of $iF_{\text{in} \rightarrow \text{in}}$, there is evidence that solid fuel combustion in household cook stoves can contribute substantially to ambient $\text{PM}_{2.5}$ concentrations in some regions (e.g., India, China) (Chafe et al., 2014).

The data given above provide inputs to predict AER, deposition, and filtration. Chen and Zhao (2011) provide a detailed review of penetration efficiency measurements and modeling strategies. While the focus of previous work has mostly been on the penetration of ambient $\text{PM}_{2.5}$ into the indoor environments, results of these studies can also be used to estimate penetration of indoor-generated particles between separated indoor zones/rooms. Tools are also available to account for evaporative losses of ammonium nitrate (Hering et al., 2007; Lunden et al., 2003), and the development of modeling tools to predict the gas-particle partitioning of SVOCs (of both indoor and outdoor origin) in indoor air is an active area of ongoing research (Hodas and Turpin, 2014; Liu et al., 2014; Weschler, 2011; Weschler and Nazaroff, 2008, 2010).

Because the availability of organic matter for sorption influences the gas-particle partitioning of SVOCs, there is the potential for the indoor formation of particles that are only present due to interactions between SVOCs of indoor and outdoor origin. For example, gas-phase SVOCs emitted indoors can sorb to indoor particulate matter of outdoor origin that has penetrated into the home (Lioy, 2006; Weschler and Nazaroff, 2008). Similarly, incoming organics from outdoors can shift from the gas-phase toward the particle phase as they sorb to particulate organic matter emitted by indoor sources (Hodas and Turpin, 2014; Naumova et al., 2003; Polidori et al., 2006; Shi and Zhao, 2012; Weschler and Nazaroff, 2008). The result is the formation of $\text{PM}_{2.5}$ that is in part, but not fully, attributable

to indoor sources. Such interactions between pollutants of indoor and outdoor origin highlight the difficulty in fully separating the contributions of indoor and outdoor $\text{PM}_{2.5}$ sources to the intake of $\text{PM}_{2.5}$.

The formation of SOA from reactions between indoor-generated VOCs and oxidants (e.g., ozone) of outdoor origin is another example of the ways in which outdoor-generated pollutants can influence the intake of $\text{PM}_{2.5}$ associated with indoor sources. Contributions of secondary particulate matter derived from well-characterized inorganic systems to outdoor iF have previously been accounted for using chemical transport models (e.g., Greco et al., 2007; Levy et al., 2003). The data and modeling tools available to include indoor secondary particulate matter (specifically, SOA) formation in estimates of indoor $\text{PM}_{2.5}$ exposures continue to improve. Waring (2014) presented a mechanistic model to calculate time-averaged indoor SOA concentrations formed as a result of the oxidation of reactive organic gases by ozone and the hydroxyl radical. Distributions of model inputs for 66 reactive organic gases relevant to the indoor environment (Turpin et al., 2007; Weisel et al., 2005) are provided in that work. In addition, a linear regression model describing SOA concentrations as a function of AER, indoor concentrations of outdoor-generated ozone and organic aerosols, indoor organic aerosol emission rates, particle and ozone deposition rates, temperature, and emission rates of reactive organic gases described the majority of variability in SOA concentrations calculated using the more complex mechanistic SOA model described above ($R^2 = 0.88$; Waring, 2014). Ji and Zhao (2015) demonstrated that the extent to which indoor SOA formation impacts indoor concentrations of $\text{PM}_{2.5}$ varies geographically, with SOA comprising 6 to 30% of indoor $\text{PM}_{2.5}$ mass for the US homes included in the Waring (2014) study, but less than 3% of $\text{PM}_{2.5}$ mass for homes in Beijing. Accounting for SOA formation indoors is an active and quickly advancing area of research and is crucial for ensuring that the full impact of specific products, activities, and processes can be taken into account in LCIA.

Discussion

Applications in life cycle impact assessment

The data provided in this review constitute a first step in addressing key questions and current challenges previously identified for the incorporation of health effects associated with indoor $\text{PM}_{2.5}$ emissions into LCIA (Fantke et al., 2015; Hellweg et al., 2009; Humbert et al., 2015). Specifically, this review allows for the characterization of a range of exposure-scenario archetypes, both in terms of indoor setting (e.g., residence, office) and in geographic location, aids in the identification of the major factors influencing $iF_{\text{in} \rightarrow \text{in}}$ and poten-

tial spatial and temporal variability in the importance of these key factors, and allows for the assessment of the level of detail and scope needed when developing exposure-scenario archetypes for use in LCIA.

In an ongoing effort, the UNEP-SETAC task force on PM_{2.5} health effects will utilize the data provided in this review to build a quantitative assessment framework for consistently combining and evaluating indoor and outdoor intake fractions from PM_{2.5} sources for application in LCIA. Complementary work is currently focusing on (1) conducting a quantitative assessment of potential variability in $iF_{in \rightarrow in}$ (e.g., across exposure scenarios and geographic regions), as well as the sensitivity of calculations of $iF_{in \rightarrow in}$ to heterogeneity in the input parameters reviewed here, (2) the evaluation of state-of-the-art modeling tools available to predict indoor and outdoor intake fractions in the context of suitability for use in LCIA, and (3) the consistent incorporation of various shapes of ERFs (Fantke et al., 2015). Together, these efforts will aid in the development of a standardized methodology by which to estimate exposures and will contribute to the effort to include PM_{2.5}-related health effects in LCIA.

Key to assessing PM_{2.5}-related health effects over the life cycle of products is the ability to evaluate the range of potential human exposure associated with a given particle emissions source. Previous work has illustrated the potential magnitude of spatial and temporal variability in $iF_{in \rightarrow in}$. Humbert et al. (2011), for example, estimates that typical values of $iF_{in \rightarrow in}$ range between approximately 10^{-3} and 10^{-2} kg intake at the population scale per kg emitted indoors. Klepeis and Nazaroff (2006) found that $iF_{in \rightarrow in}$ for environmental tobacco smoke varied between 6.6×10^{-4} and 2.6×10^{-3} kg intake per kg emitted within a single simulated home depending on multiple factors including home ventilation conditions and occupant activity patterns. Thus, while a single recommended value meant to characterize a needed modeling parameter is valuable for providing an estimate of the magnitude of $iF_{in \rightarrow in}$ (e.g., a single AER value meant to represent typical housing the USA), distributions or ranges describing these input parameters are crucial. Such distributions allow for the evaluation of the central tendencies of $iF_{in \rightarrow in}$, as well as the extremes, thereby acknowledging the variability in population exposure patterns, housing aspects, and indoor air chemistry. By aggregating the results of multiple studies, the present review provides a broader picture of the range of potential values for a given parameter influencing indoor concentrations of PM_{2.5} and allows for the consideration of a range of archetypal indoor environments. It is important to note that these values vary temporally and spatially with multiple factors, as discussed in the individual sections above, and parameters are not available to describe all exposure scenarios and geographic regions. Thus, understanding the full range

of input parameters also allows for the consideration of uncertainty in $iF_{in \rightarrow in}$ for PM_{2.5}.

Depending on the design of the selected modeling framework, not all of the factors potentially contributing to variability in $iF_{in \rightarrow in}$ will necessarily be considered in LCIA. For example, Hellweg et al. (2009) suggested that the representation of the indoor environment as a single, well-mixed compartment provides the most effective way to incorporate indoor PM_{2.5} exposures into LCIA. On the other hand, in regards to assessing exposure to individual VOCs from cleaning products, Earnest and Corsi (2013) propose the use of a two-zone model in which the near-person/near-source region and the rest of the indoor environment are treated as discrete zones. LCIA often follows approaches based on archetypes to account for differences in exposure scenarios or geographic regions. Thus, the parameters that will be of the greatest importance are those that account for geographic variability in more general housing and building characteristics (e.g., volume, whole-building air exchange and ventilation), indoor environment occupancy, and the prevalence of specific indoor sources (e.g., cooking and heating appliances). Parameters that provide a higher level of detail (e.g., activity-specific breathing rates, local-scale flows), however, will be valuable to higher tier assessments of indoor air quality and epidemiologic studies that aim to characterize indoor PM_{2.5} exposures for specific conditions in a well-characterized environment.

Remaining limitations and recommendations for future research

One contributor to limitations in the availability and scope of data like those reviewed here is the fact that the studies carried out to collect the data are expensive and work intensive. As a result, they tend to be carried out in infrequent, intensive campaigns. As noted above, for example, many AER studies are not representative of the full range of housing stock, even for the nations or cities in which they were carried out. Values are more limited or non-existent in some developing countries and are biased toward USA and European studies. We suggest that there is a need for studies on AER in developing countries, particularly in rural regions where biomass is used for cooking in homes.

Another issue constraining the representativeness of the data is the potential for changes with time. While some values are not expected to vary temporally (i.e., *IR*, although the activity levels driving them may change), others change on timescales faster than the studies characterizing them are carried out. Bongaarts (2001), for example, noted the tendency for household size to converge toward the nuclear family in rapidly industrializing and urbanizing regions. Similarly, there is the potential for changes in human activity patterns with increased access to media, suggesting a need for

updated human activity pattern data. Housing construction practices change with advancing technology and materials development, as well as with recent pushes toward energy efficiency. Urban growth (e.g., Schneider and Woodcock, 2008; Seto and Fragkias, 2005; Xiao et al., 2006) may make the lack of data characterizing AERs in apartments and multifamily residences a major issue in both developing and developed countries. New techniques utilizing 3D imaging sensors to evaluate building/room size and leakage characteristics show promise in increasing data availability for leaky buildings (e.g., in developing countries), airtight, energy-efficient buildings, and multifamily residences (Gong and Caldas, 2008) and should be a consideration in future work in this area. Finally, while the principles driving pollutant dynamics will not change with time, emission rates, particle size distributions, and particle composition may change with technology. Cynthia et al. (2008), for example, reported a 35% decrease in PM_{2.5} exposures with the introduction of a higher efficiency cook stove in an intervention study in rural Mexico. As a result of these ever-changing factors, a continued effort to undertake such studies and to expand their temporal and spatial scope is key to ensuring that the impacts associated with specific products and emission sources can be fully assessed in the context of LCIA.

We also recommend that future efforts focus on a number of key research areas. First, there is a need for a more widespread and detailed characterization of inter- and intrazonal airflows and the factors that influence them for a range of residence types, commercial buildings, and occupational settings to derive useful information for higher tier assessments of indoor air quality. Such characterizations would be useful in addressing proximity-to-source issues. Of particular importance may be the development of a set of archetypal building layouts that describe a range of building types, so that these highly variable flows can be modeled for a given exposure scenario with tools such as COMIS and CONTAM. For applications in LCIA, a simple two-zone model might be more suitable as more complex approaches might lack data and consistency across indoor and outdoor emission situations. As noted above, there are large geographic differences in the heating and cooling systems present in households and other indoor environments on a global scale. Documenting these differences and the related impacts on indoor particle dynamics is an important area of future work. Finally, there is a need for more research

aimed at obtaining a thorough understanding of interactions between indoor- and outdoor-generated pollutants and the formation of SOA in indoor air. Key to this is the development of accurate simplified models that can easily be applied in LCIA. The regression model developed by Waring (2014) to predict indoor SOA formation based on a small number of key parameters provides an example of the type of modeling tools that will advance predictions of $iF_{in \rightarrow in}$ for PM_{2.5} in this context.

Conclusions

The present paper reviews and compiles the results of studies exploring the main factors influencing indoor PM_{2.5} concentrations and associated $iF_{in \rightarrow in}$, with an emphasis on primary indoor PM_{2.5} emissions. Specifically, we focus on factors related to building characteristics, occupant characteristics and behaviors, and pollutant properties and dynamics. The key studies and data sources discussed herein comprise a tool kit of exposure-modeling parameters that can be used to estimate the central tendencies and potential ranges of $iF_{in \rightarrow in}$. A follow-up effort will utilize the data provided in the present review to build a framework to consistently integrate indoor and outdoor exposures to PM_{2.5} emitted by indoor and outdoor sources. Combined, the present review and the follow-up work contribute to the effort to consistently include PM_{2.5}-derived health effects in LCIA. Continued efforts to characterize the factors influencing indoor PM_{2.5} concentrations will ensure that impacts associated with specific products and emission sources can be fully assessed in LCIA and other comparative human exposure and impact assessment frameworks.

Acknowledgements

This work was supported by the UNEP/SETAC Life Cycle Initiative. Natasha Hodas was funded by National Science Foundation award no. 1433246.

Supporting Information

Additional Supporting Information may be found in the online version of this article:

Data S1. All data characterizing building-specific, human-specific, and pollutant-specific factors that influence indoor inhalation intake fraction of fine particulate matter considered in this review are provided in tabular form in the SI.

References

- Abt, E., Suh, H.H., Catalano, P. and Koutrakis, P. (2000) Relative contribution of outdoor and indoor particles sources to indoor concentrations, *Environ. Sci. Technol.*, **34**, 3579–3587.
- Adams, W.C. (1993) Measurement of breathing rate and volume in routinely performed daily activities. Contract No.

- A033-205 Air Resources Board, Sacramento, CA.
- Aelenei, D., Azevedo, S., Viegas, J., Mendes, A., Papoila, A.L., Cano, M., Martins, P. and Neuparth, N. (2013) Environment and health in children day care centers in Portugal. Results from phase II on the ventilation characteristics of 16 schools, 11th Annual REHVA World Congress and 8th International Conference on IAQVEC.
- Afshari, A., Matson, U. and Ekberg, L.E. (2005) Characterization of indoor sources of fine and ultrafine particles: a study conducted in a full-scale chamber, *Indoor Air*, **15**, 141–150.
- Allen, R.W., Adar, S.D., Avol, E., Cohen, M., Curl, C.L., Larson, T., Liu, L.-J.S., Sheppard, L. and Kaufman, J.D. (2012) Modeling the residential infiltration of outdoor PM_{2.5} in the Multi-Ethnic Study of Atherosclerosis and Air Pollution (MESA Air), *Environ. Health Perspect.*, **120**, 824–830.
- Apple, J., Vincete, R., Yarberry, A., Lohse, N., Mills, E., Jacobson, A. and Poppendieck, D. (2010) Characterization of particulate matter size distributions and indoor concentrations from kerosene and diesel lamps, *Indoor Air*, **20**, 399–411.
- Armendáriz-Arnez, C., Edwards, R.D., Johnson, M., Rosas, I.A., Espinosa, F. and Masera, O.R. (2010) Indoor particle size distributions in homes with open fires and improved Patsari cook stoves, *Atmos. Environ.*, **44**, 2881–2886.
- Asikainen, A., Hanninen, O., Brelih, N., Bischof, W., Hartmann, T., Carrer, P. and Wargocki, P. (2013) The proportion of residences in European countries with ventilation rates below the regulation base limit value, *Int. J. Vent.*, **12**, 129–134.
- Australian Centre for Human Health Risk Assessment (2012) Office of Health Protection of the Australian Government Department of Health Australian Exposure Factor Guide, <http://www.health.gov.au/internet/main/publishing.nsf/Content/health-pubhlth-publicat-enviro.htm>.
- Azimi, P., Zhao, D. and Stephens, B. (2014) Estimates of HVAC filtration efficiency for fine and ultrafine particles of outdoor origin, *Atmos. Environ.*, **98**, 337–346.
- Baek, S., Kim, Y. and Perry, R. (1997) Indoor air quality in homes, offices and restaurants in Korean urban areas – indoor-outdoor relationships, *Atmos. Environ.*, **3**, 529–544.
- Bari, M.A., MacNeill, M., Kindziarski, W.B., Wallace, L., Héroux, M.-È. and Wheeler, A.J. (2014) Predictors of coarse particulate matter and associated endotoxin concentrations in residential environments, *Atmos. Environ.*, **92**, 221–230.
- Bennett, D.H. and Furtaw, E.J. Jr (2004) Fugacity-based indoor residential pesticide fate model, *Environ. Sci. Technol.*, **38**, 2142–2152.
- Bennett, D.H., McKone, T.H., Evans, J.S., Nazaroff, W.W., Margni, M.D., Jolliet, O. and Smith, K.R. (2002) Defining iF, *Environ. Sci. Technol.*, **36**, 206–211.
- Bennett, D.H., Fisk, W., Apte, M.G., Wu, X., Trout, A., Faulkner, D. and Sullivan, D. (2012) Ventilation, temperature, and HVAC characteristics in small and medium commercial buildings in California, *Indoor Air*, **22**, 309–320.
- Bongaarts, J. (2001) Household size and composition in the developing world in the 1990s, *Popul. Stud.*, **55**, 263–279.
- Bonjour, S., Adair-Rohani, H., Wolf, J., Bruce, N.G., Metha, S., Pruss-Ustun, A., Lahiff, M., Rehfuess, E.A., Mishra, V. and Smith, K.R. (2013) Solid fuel use for household cooking: country and regional estimates for 1980-2010, *Environ. Health Perspect.*, **121**, 784–790.
- Breen, M.S., Burke, J.M., Batterman, S.A., Vette, A.F., Godwin, C., Croghan, C.W., Schultz, B.D. and Long, T.C. (2014) Modeling spatial and temporal variability of residential air exchange rates for the Near-Road Exposures and Effects of Urban Air Pollutants Study (NEXUS), *Int. J. Environ. Res. Public Health*, **11**, 11481–11504.
- Brelih, N. and Seppänen, O. (2011) Ventilation rates and IAQ in European standards and national regulations, In: *The Proceedings of the 32nd AIVC Conference and 1st TightVent Conference in Brussels*, 12–13 October 2011. Available online: www.aivc.org.
- Burke, J.M., Zufall, M.J. and Ozkaynak, H. (2001) A population exposure model for particulate matter: case study results for PM_{2.5} in Philadelphia, PA, *J. Expo. Anal. Environ. Epidemiol.*, **11**, 470–489.
- Canha, N., Almelda, S.M., Freitas, M.C., Taubel, M. and Hanninen, O. (2013) Winter ventilation rates at primary schools: comparison between Portugal and Finland, *J. Toxicol. Environ. Health*, **76**, 400–408.
- Chafe, Z., Brauer, M., Klimont, Z., Van Dingenen, R., Mehta, S., Rao, S., Riahi, K., Dentener, F. and Smith, K.R. (2014) Household cooking with solid fuels contributes to ambient PM_{2.5} air pollution and the burden of disease, *Environ. Health Perspect.*, **122**, 1314–1320.
- Chan, W.R., Nazaroff, W.W., Price, P.N., Sohn, M.D. and Gadgil, A.J. (2005) Analyzing a database of residential air leakage in the U.S., *Atmos. Environ.*, **39**, 3445–3455.
- Chan, W.R., Cohn, S., Sidheswaran, M., Sullivan, D.P. and Fisk, W.J. (2014) Contaminant levels, source strengths, and ventilation rates in California retail stores, *Indoor Air*, **25**, 381–392.
- Chao, C. (2001) Comparison between indoor and outdoor contaminant levels in residential buildings from passive sampler study, *Build. Environ.*, **36**, 999–1007.
- Chen, C. and Zhao, B. (2011) Review of relationship between indoor and outdoor particles: I/O ratio, infiltration factor and penetration factor, *Atmos. Environ.*, **45**, 275–288.
- Chen, C., Zhao, B. and Weschler, C.J. (2012) Assessing the influence of indoor exposure to “outdoor ozone” on the relationship between ozone and short-term mortality in U.S. communities, *Environ. Health Perspect.*, **120**, 235–240.
- Clark, N.A., Allen, R.W., Hystad, P., Wallace, L., Dell, S.D., Foty, R., Dabek-Zlotorzynska, E., Evans, G. and Wheeler, A.J. (2010) Exploring variation and predictors of residential fine particulate matter infiltration, *Int. J. Environ. Res. Public Health*, **7**, 3211–3224.
- Coley, D.A. and Beisteiner, A. (2002) Carbon dioxide levels and ventilation rates in schools, *Int. J. Vent.*, **1**, 45–52.
- Cynthia, A.A., Edwards, R.D., Johnson, M., Zuk, M., Rojas, L., Jimenez, R.D., Riojas-Rodriguez, H. and Masera, O. (2008) Reduction in personal exposures to particulate matter and carbon monoxide as a result of the installation of a Patsari improved cook stove in Michoacan Mexico, *Indoor Air*, **18**, 93–105.
- Diapouli, E., Chaloulakou, A. and Koutrakis, P. (2013) Estimating the concentration of indoor particles of outdoor origin: a review, *J. Air Waste Manag. Assoc.*, **63**, 1113–1129.
- Dimitroulopoulou, C. (2012) Ventilation in European dwellings: a review, *Build. Environ.*, **47**, 109–125.
- Dimitroulopoulou, C. and Bartzis, J. (2013) Ventilation rates in European office buildings: a review, *Indoor Built Environ.*, **23**, 5–25.
- D’Oca, S. and Hong, T. (2014) A data-mining approach to discover patterns of window opening and closing behavior in offices, *Build. Environ.*, **82**, 726–739.
- Drescher, A.C., Lobascio, C., Gadgil, A.J. and Nazaroff, W.W. (1995) Mixing of a point-source indoor pollutant by forced convection, *Indoor Air*, **5**, 204–214.
- Du, L., Batterman, S., Godwin, C., Chin, J.-Y., Parker, E., Breen, M., Brakefield, W., Robins, T. and Lewis, T. (2012) Air change rates and interzonal flows in residences, and the need for multi-zone models for exposure and health analyses, *Int. J. Environ. Res. Public Health*, **9**, 4639–4661.
- Dutton, S.M., Mendell, M.J., Chan, W.R., Barrios, M., Sidheswaran, M.A., Sullivan, D.P., Eliseeva, E.A. and Fisk, W.J. (2015) Evaluation of the indoor air quality minimum ventilation rate procedure

- for use in California retail buildings, *Indoor Air*, **25**, 93–104.
- Earnest, C.M. and Corsi, R.L. (2013) Inhalation exposure to cleaning products: application of a two-zone model, *J. Occup. Environ. Hyg.*, **10**, 328–335.
- Edwards, R., Karnani, S., Fisher, E.M., Johnson, M., Naeher, L., Smith, K.R. and Morawska, L. (2014) WHO IAQ Guidelines: household fuel combustion – Review 2: Emissions, http://www.who.int/indoorair/guidelines/hhfc/Review_2.pdf.
- El Orchi, Z., Stephens, B. and Waring, M.S. (2014) Predictions and determinants of size-resolved particle infiltration factors in single-family homes in the U.S., *Build. Environ.*, **74**, 106–118.
- Emmerich, S.J. (2001) Validation of multi-zone IAQ modeling of residential-scale buildings: a review, *ASHRAE Trans.*, **107**, 619–628.
- Emmerich, S.J. and Crum, J. (2006) Simulated performance of natural and hybrid ventilation systems in an office building, *HVAC & R Res.*, **12**, 975–1004.
- Emmerich, S.J., Howard-Reed, C. and Nabinger, S.J. (2004) Validation of multi-zone IAQ model predictions for tracer gas in a townhouse, *Build. Serv. Eng. Res. Technol.*, **25**, 305–316.
- Eurostat, 2014. Average household size (source: European Union Statistics on Income and Living Conditions). European Commission, accessed: 7 June 2014.
- Evans, G.J., Peers, A. and Sabaliauskas, K. (2008) Particle dose estimation from frying in residential settings, *Indoor Air*, **18**, 499–510.
- Fabi, V., Andersen, R.V., Corgnati, S. and Olesen, B.W. (2012) Occupants' window opening behavior: a literature review of factors influencing occupant behaviors and models, *Build. Environ.*, **58**, 188–198.
- Fadeyi, M.O., Weschler, C.J. and Tham, K.W. (2009) The impact of recirculation, ventilation and filters on secondary organic aerosols generated by indoor chemistry, *Atmos. Environ.*, **43**, 3538–3547.
- Fantke, P., Jolliet, O., Apte, J.S., Cohen, A.J., Evans, J.S., Hänninen, O.O., Hurley, F., Jantunen, M.J., Jerrett, M., Levy, J.I., Loh, M.M., Marshall, J.D., Miller, B.G., Preiss, P., Spadaro, J.V., Tainio, M., Tuomisto, J.T., Weschler, C.J. and McKone, T.E. (2015) Health effects of fine particulate matter in life cycle impact assessment: conclusions from the Basel guidance workshop, *Int. J. Life Cycle Assess.*, **20**, 276–288.
- Ferro, A.R., Kopperud, R.J. and Hildemann, L.M. (2004) Source strengths for indoor human activities that resuspend particulate matter, *Environ. Sci. Technol.*, **38**, 1759–1764.
- Feustel, H.E. (1998) COMIS – An international multizone air-flow and contaminant transport model, *Energy Build.*, **30**, 3–18.
- Finlayson-Pitts, B.J. and Pitts, J.N. Jr (2000) *Chemistry of the Upper and Lower Atmosphere*, San Diego, CA, Academic Press.
- Fisher, K. and Tucker, J., Contributions from Evrim Altintas, Matthew Bennett, Antony Jahandar, Jiweon Jun, and other members of the Time Use Team (2013) Technical Details of Time Use Studies. Last updated 15 July 2013. Centre for Time Use Research, University of Oxford, United Kingdom. <http://www.timeuse.org/information/studies/>
- Gadgil, A.J., Lobscheid, C., Abadie, M.O. and Finlayson, E.U. (2003) Indoor pollutant mixing time in an isothermal closed room: an investigation using CFD, *Atmos. Environ.*, **37**, 5577–5586.
- Gehin, E., Ramalho, O. and Kirchner, S. (2008) Size distribution and emission rate measurement of fine and ultrafine particle from indoor human activities, *Atmos. Environ.*, **42**, 8341–8352.
- Gong, J. and Caldas, H. (2008) Data processing for real-time construction site spatial modeling, *Autom. Constr.*, **17**, 526–535.
- Gózeński, R., Symański, M., Górka, A. and Mróz, T. (2014) Airtightness of buildings in Poland, *Int. J. Ventilation*, **12**, 391–400.
- Greco, S.L., Wilson, A.M., Spengler, J.D. and Levy, J.I. (2007) Spatial patterns of mobile source particulate matter emissions-to-exposure relationships across the United States, *Atmos. Environ.*, **41**, 1011–1025.
- Guo, Z., Jetter, J.J. and McBrien, J.A. (2004) Rates of polycyclic aromatic hydrocarbon emissions from incense, *Bull. Environ. Contam. Toxicol.*, **72**, 186–193.
- Guo, H., Morawska, L., He, C. and Gilbert, D. (2008) Impact of ventilation scenario on air exchange rates and on indoor particle number concentrations in an air-conditioned classroom, *Atmos. Environ.*, **42**, 757–768.
- Gurunathan, S., Robson, M., Freeman, N., Buckley, B., Roy, A., Meyer, R., Bukowski, J. and Lioy, P.J. (1998) Accumulation of chlorpyrifos on residential surfaces and toys accessible to children, *Environ. Health Perspect.*, **106**, 9–16.
- Haas, A., Weber, A., Dorer, V., Keilholz, W. and Pelletret, R. (2002) COMIS v3.1 simulation environment for multizone air flow and pollutant transport modeling, *Energy Build.*, **34**, 873–882.
- Habib, G., Venkataraman, C., Bond, T.C. and Schauer, J.J. (2008) Chemical, microphysical, and optical properties of primary particles from the combustion of biomass fuels, *Environ. Sci. Technol.*, **42**, 8829–8834.
- Hanley, J.T., Ensor, D.S., Smith, D.D. and Sparks, L.E. (1994) Fractional aerosol filtration efficiency of in-duct ventilation air cleaners, *Indoor Air*, **3**, 169–178.
- Hänninen, O., Hoek, G., Mallone, S., Chellini, E., Katsouyanni, K., Gariazzo, C., Cattani, G., Marconi, A., Molnar, P., Bellander, T. and Jantunen, M. (2011) Seasonal patterns of outdoor PM infiltration into indoor environments: review and meta-analysis of available studies from different climatological zones in Europe, *Air Qual. Atmos. Health*, **4**, 221–233.
- Hänninen, O., Sorjamaa, R., Lipponen, P., Cyrys, J., Lanki, T. and Pekkanen, J. (2013) Aerosol-based modeling of infiltration of ambient PM_{2.5} and evaluation against population-based measurements in homes in Helsinki, Finland, *J. Aerosol Sci.*, **66**, 111–122.
- Hays, M.D., Smith, N.D., Kinsey, J., Dong, Y. and Kariher, P. (2003) Polycyclic aromatic hydrocarbon size distributions in aerosols from appliances of residential wood combustion as determined by direct thermal desorption—GC/MS, *J. Aerosol Sci.*, **34**, 1061–1084.
- He, C., Morawska, L., Hitchins, J. and Gilbert, D. (2004) Contribution from indoor sources to particle number and mass concentrations in residential houses, *Atmos. Environ.*, **38**, 3405–3415.
- He, C., Morawska, L. and Gilbert, D. (2005) Particle deposition rates in residential houses, *Atmos. Environ.*, **39**, 3891–3899.
- He, C., Morawska, L. and Taplin, L. (2007) Particle emission characteristics of office printers, *Environ. Sci. Technol.*, **41**, 6039–6045.
- Hellweg, S., Demou, E., Bruzzi, R., Meijer, A., Rosenbaum, R.K., Huijbregts, M.A.J. and McKone, T.E. (2009) Integrating human indoor air pollutant exposure within life cycle impact assessment, *Environ. Sci. Technol.*, **43**, 1670–1679.
- Hering, S.V., Lunden, M.M., Thatcher, T.L., Kirchstetter, T.W. and Brown, N.J. (2007) Using regional data and building leakage to assess indoor concentrations of particles of outdoor origin, *Aerosol Sci. Technol.*, **41**, 639–654.
- Hodas, N. and Turpin, B.J. (2014) Shifts in the gas-particle partitioning of ambient organics with transport into the indoor environment, *Aerosol Sci. Technol.*, **48**, 271–281.
- Hodas, N., Meng, Q., Lunden, M.M., Rich, D.Q., Ozkaynak, H., Baxter, L.K., Zhang, Q. and Turpin, B.J. (2012) Variability in the fraction of ambient fine particulate matter found indoors and observed heterogeneity in health effect estimates, *J. Expo. Sci. Environ. Epidemiol.*, **22**, 448–454.

- Hodas, N., Meng, Q., Lunden, M.M. and Turpin, B.J. (2014) Toward refined estimates of ambient PM_{2.5} exposure: evaluation of a physical outdoor-to-indoor transport model, *Atmos. Environ.*, **83**, 229–236.
- Howard-Reed, C., Wallace, L.A. and Emmerich, S.J. (2003) Effect of ventilation systems and air filters on decay rates of particles produced by indoor sources in an occupied townhouse, *Atmos. Environ.*, **37**, 5295–5306.
- Huang, K., Feng, G., Li, H. and Yu, S. (2014) Opening window issue of residential buildings in winter in north China: a case study in Shenyang, *Energy Build.*, **84**, 567–574.
- Humbert, S., Marshall, J.D., Shaked, S., Spadarp, J.V., Nishioka, Y., Preiss, P., McKone, T.E., Horvath, A. and Jolliet, O. (2011) IF for particulate matter: recommendations for life cycle impact assessment, *Environ. Sci. Technol.*, **45**, 4808–4816.
- Humbert, S., Fantke, P. and Jolliet, O. (2015) Particulate matter formation, In: Hauschild, M. and Huijbregts, M.A.J. (eds) *Life Cycle Impact Assessment*, Dordrecht, Springer Press, 97–113.
- Hussein, T., Glystos, T., Ondracek, J., Dohanyosova, P., Zdimal, V., Hameri, K., Lazaridis, M., Smolik, J. and Kulmala, M. (2006) Particle size characterization and emission rate during indoor activities in a house, *Atmos. Environ.*, **40**, 4285–4307.
- Ilacqua, V., Hänninen, O., Kuenzli, N. and Jantunen, M.F. (2007) IF distributions for indoor VOC sources in five European cities, *Indoor Air*, **17**, 372–383.
- Iwashita, G. and Akasaka, H. (1997) The effects of human behavior on natural ventilation rate and indoor air environment in summer – a field study in southern Japan, *Energy Build.*, **25**, 195–205.
- Jang, J.-Y., Kim, S.-Y., Kim, S.-J., Lee, K.-E., Cheong, H.-K., Kim, E.-H., Choi, K.-H. and Kim, Y.-H. (2014a) General factors of the Korean exposure factors handbook, *J. Prev. Med. Public Health*, **47**, 7–17.
- Jang, J.-Y., Jo, S.-N., Kim, S.-Y., Lee, K.-E., Choi, K.-H. and Kim, Y.H. (2014b) Activity factors of the Korean exposure factors handbook, *J. Prev. Med. Public Health*, **47**, 27–35.
- Jetter, J.J. and Kariher, P. (2009) Solid-fuel household cook stoves: characterization of performance and emissions, *Biomass Bioenergy*, **33**, 294–305.
- Jetter, J.J., Guo, Z., McBrian, J.A. and Flynn, M.R. (2002) Characterization of emissions from burning incense, *Sci. Total Environ.*, **295**, 51–67.
- Jetter, J., Zhao, Y., Smith, K.R., Khan, B., Yelverton, T., DeCarlo, P. and Hays, M.D. (2012) Pollutant emissions and energy efficiency under controlled conditions for household biomass cookstoves and implications for metrics useful in setting international test standards, *Environ. Sci. Technol.*, **46**, 10827–10834.
- Ji, W. and Zhao, B. (2015) Contribution of outdoor-originating particles, indoor-emitted particles and indoor secondary organic aerosols (SOA) to residential indoor PM_{2.5} concentration: a model-based estimation, *Build. Environ.*, **90**, 196–205.
- Johnson, T. and Long, T. (2005) Determining the frequency of open windows in residences: a pilot study in Durham, North Carolina during varying temperature conditions, *J. Expo. Anal. Environ. Epidemiol.*, **15**, 329–349.
- Jolliet, O., Frischknecht, R., Bare, J., Boulay, A.-M., Bulle, C., Fantke, P., Gheewala, S., Hauschild, M., Itsubo, N., Margni, M., McKone, T., Mila y Canals, L., Postuma, L., Prado, V., Ridoutt, B., Sonneman, G., Rosenbaum, R., Seager, T., Struis, J., van Zelm, R., Vigon, B. and Weisbrod, A. (2014) Global guidance on environmental life cycle impact assessment indicators: findings of the scoping phase, *Int. J. Life Cycle Assess.*, **19**, 962–967.
- Kearny, J., Wallace, L., MacNeill, M., Héroux, M.-E. and Kindzierski, W. (2014) Residential infiltration of fine and ultrafine particles in Edmonton, *Atmos. Environ.*, **94**, 793–805.
- Kleeman, M.J., Schauer, J.J. and Cass, G.R. (1999) Size and composition distribution of fine particulate matter emitted from wood burning, meat charbroiling, and cigarettes, *Environ. Sci. Technol.*, **33**, 3516–3523.
- Klepeis, N.E. (2004) Using computer simulation to explore multi-compartment effects and mitigation strategies for residential exposure to secondhand tobacco smoke, Dissertation (Ph.D.), Environmental Health Sciences, Berkeley, University of California.
- Klepeis, N.E. and Nazaroff, W.W. (2006) Modeling residential exposure to secondhand tobacco smoke, *Atmos. Environ.*, **40**, 4393–4407.
- Klepeis, N.E., Nelson, W.C., Ott, W.R., Robinson, J.P., Tsang, A.M., Switzer, P., Behar, J.V., Stephen, C.H. and Engelmann, W.H. (2001) The National Human Activity Pattern Survey (NHAPS): a resource for assessing exposure to environmental pollutants, *J. Expo. Anal. Environ. Epidemiol.*, **11**, 231–252.
- Lai, A.C.K. (2002) Particle deposition indoors: a review, *Indoor Air*, **12**, 211–214.
- Lai, A.C.K., Thatcher, T.L. and Nazaroff, W.W. (2000) Inhalation transfer factors for air pollution health risk assessment, *J. Air Waste Manag. Assoc.*, **50**, 1688–1699.
- Lee, S.-C. and Wang, B. (2004) Characteristics of emissions of air pollutants from burning of incense in a large environmental chamber, *Atmos. Environ.*, **38**, 941–951.
- Levie, D., de Kluizenaar, Y., Hoes-van Offelen, E.C.M., Hofstetter, H., Janssen, S.A., Spiekman, M.E. and Koene, F.G.H. (2014) Determinants of ventilation behavior in naturally ventilated dwellings: identification and quantification of relationships, *Build. Environ.*, **82**, 388–399.
- Levy, J.I., Wilson, A.M., Evans, J.S. and Spengler, J.D. (2003) Estimation of primary and secondary particulate matter intake fractions for power plants in Georgia, *Environ. Sci. Technol.*, **37**, 5528–5536.
- Li, W. and Hopke, P.K. (1993) Initial size distributions and hygroscopicity of indoor combustion aerosol particles, *Aerosol Sci. Technol.*, **19**, 305–316.
- Li, Y. and Li, X. (2015) Natural ventilation potential of high-rise residential buildings in northern China using coupling thermal and airflow simulations, *Build. Simul.*, **8**, 51–64.
- Li, Y., Lueng, G.M., Tang, J.W., Yang, X., Chao, C.Y.H., Lin, J.Z., Lu, J.W., Nielsen, P.V., Niu, J., Qian, H., Sleigh, A.C., Su, H.J.J., Sundell, J., Wong, T.W. and Yeun, P.L. (2007) Role of ventilation in airborne transmission of infectious agents in the built environment – a multidisciplinary systematic review, *Indoor Air*, **17**, 2–18.
- Li, N., Li, J., Fan, R. and Jia, H. (2015) Probability of occupant operation of windows during transition seasons in office buildings, *Renewable Energy*, **73**, 84–91.
- Liang, C. and Pankow, J.F. (1996) Gas/particle partitioning of organic compounds to environmental tobacco smoke: partitioning coefficient measurements by desorption and comparison to urban particulate material, *Environ. Sci. Technol.*, **30**, 2800–2805.
- Lioy, P. (2006) Employing dynamical and chemical processes for contaminant mixtures outdoors to the indoor environment: the implications for total human exposure analysis and prevention, *J. Expo. Sci. Environ. Epidemiol.*, **16**, 207–224.
- Litiu, A. (2012) Ventilation system types in some EU countries, *REHVA J.*, 18–23.
- Liu, W., Zhang, J., Hashim, J.H., Jalaludin, J., Hashim, Z. and Goldstein, B.D. (2003) Mosquito coil emissions and health implications, *Environ. Health Perspect.*, **111**, 1454–1460.
- Liu, C., Zhang, Y. and Weschler, C.J. (2014) The impact of mass transfer limitations on size distributions of particle associated SVOCs in outdoor and indoor environments, *Sci. Total Environ.*, **497**, 401–411.

- Long, C.M., Suh, H.H. and Koutrakis, P. (2000) Characterization of indoor particle sources using continuous mass and size monitors, *J. Air Waste Manag. Assoc.*, **50**, 1236–1250.
- Long, C.M., Suh, H.H., Catalano, P.J. and Koutrakis, P. (2001) Using time- and size-resolved particulate data to quantify indoor penetration and deposition behavior, *Environ. Sci. Technol.*, **35**, 2089–2099.
- L'Orange, C., Leith, D., Volckens, J. and DeFoort, M. (2015) A quantitative model of cookstove variability and field performance: implications for sample size, *Bio-mass Bioenergy*, **72**, 233–241.
- Lunden, M.M., Revzan, K.L., Fischer, M.L., Thatcher, T.L., Littlejohn, D., Hering, S.V. and Brown, N.J. (2003) The transformation of outdoor ammonium nitrate aerosols in the indoor environment, *Atmos. Environ.*, **37**, 5633–5644.
- Lung, S.-C.C. and Hu, S.-C. (2003) Generation rates and emission factors of particulate matter and particle-bound polycyclic aromatic hydrocarbons of incense sticks, *Chemosphere*, **50**, 673–679.
- MacNeill, M., Wallace, L., Kearney, J., Allen, R.W., Van Ryswyk, K., Judek, S., Xu, X. and Wheeler, A. (2012) Factors influencing variability in the infiltration of PM_{2.5} mass and its components, *Atmos. Environ.*, **61**, 518–532.
- MacNeill, M., Kearney, J., Wallace, L., Gibson, M., Heroux, M.E., Kuchta, J., Guernsey, J.R. and Wheeler, A.J. (2014) Quantifying the contribution of ambient and indoor-generated fine particles to Indoor Air in residential environments, *Indoor Air*, **24**, 362–375.
- Marr, D., Mason, M., Mosley, R. and Liu, X. (2012) The influence of opening windows and doors on the natural ventilation rate of a residential building, *HVAC & R. Res.*, **18**, 195–203.
- Matz, C., Stieb, D., Davis, K., Egyed, M., Rose, A., Chou, B. and Brion, O. (2014) Effects of age, season, gender and urban-rural status on time-activity: canadian human activity pattern survey 2 (CHAPS 2), *Int. J. Environ. Res. Public Health*, **11**, 2108–2124.
- McGrath, J.A., Byrne, M.A., Ashmore, M.R., Terry, A.C. and Dimitroulopoulou, C. (2014) A simulation study of the changes in PM_{2.5} concentrations due to interzonal airflow variations caused by internal door opening patterns, *Atmos. Environ.*, **87**, 183–188.
- McNeil, S., Quaglia, L., Bassett, M., Overton, G. and Plagmann, M. (2012) A survey of airtightness and ventilation rates in post 1994 NZ homes, In *AIVC 33rd conference: optimising ventilative cooling and airtightness for [nearly] zero-energy buildings, IAQ and comfort*.
- Meng, Q.Y., Turpin, B.J., Polidori, A., Lee, J.H., Weisel, C.P., Morandi, M., Winer, A. and Zhang, J. (2005) PM_{2.5} of ambient origin: estimates and exposure errors relevant to PM epidemiology, *Environ. Sci. Technol.*, **39**, 5105–5112.
- Meng, Q.Y., Turpin, B.J., Lee, J.H., Polidori, A., Weisel, C.P., Morandi, M., Colome, S., Zhang, J., Stock, T. and Winer, A. (2007) How does infiltration behavior modify the composition of ambient PM_{2.5} in indoor spaces? An analysis of RIOPA data, *Environ. Sci. Technol.*, **41**, 7315–7321.
- Mi, Y.-H., Norback, D., Tao, J., Mi, Y.-L. and Ferm, M. (2006) Current asthma and respiratory symptoms among pupils in Shanghai, China: influence of building ventilation, nitrogen dioxide, ozone, and formaldehyde in classrooms, *Indoor Air*, **16**, 454–464.
- Naumova, Y.Y., Offenberg, J.H., Eisenreich, S.J., Meng, Q.Y., Polidori, A., Turpin, B.J., Weisel, C.P., Morandi, M.T., Colome, S.D., Stock, T.H., Winer, A.M., Alimokhtari, S., Kwon, J., Maberti, S., Shendell, D., Jones, J. and Farrar, C. (2003) Gas/particle distribution of polycyclic aromatic hydrocarbons in coupled indoor/outdoor atmospheres, *Atmos. Environ.*, **37**, 703–719.
- Nazaroff, W.W. (2004) Indoor particle dynamics, *Indoor Air*, **14**, 175–183.
- Nazaroff, W.W. (2008) Inhalation intake fraction of pollutants from episodic indoor emissions, *Build. Environ.*, **43**, 269–277.
- Nazaroff, W.W. and Weschler, C.J. (2004) Cleaning products and air fresheners: exposure to primary and secondary air pollutants, *Atmos. Environ.*, **38**, 2841–2865.
- Ogulei, D., Hopke, P.K. and Wallace, L.A. (2006) Analysis of indoor particle size distributions in an occupied townhouse using positive matrix factorization, *Indoor Air*, **16**, 204–215.
- Olson, D.A. and Burke, J.M. (2006) Distributions of PM_{2.5} source strengths for cooking from the Research Triangle Park particulate matter panel study, *Environ. Sci. Technol.*, **40**, 163–169.
- Orru, H., Mikola, A., Upan, M. and Koiv, T.-A. (2014) Variation of indoor/outdoor particulates in Tallinn, Estonia – The role of ventilation, heating systems and lifestyle, *J. Environ. Pollut. Hum. Health*, **2**, 52–57.
- Ott, W., Wallace, L. and Mage, D. (2000) Predicting particulate (PM₁₀) personal exposure distributions using a random component superposition statistical model, *J. Air Waste Manag. Assoc.*, **50**, 1390–1406.
- Ozkaynak, H., Xue, J., Weker, R., Butler, D., Koutrakis, P. and Spengler, J. (1997) *The Particle Team (PTEAM) study: Analysis of the Data, (Project Summary)*, Washington, DC, U.S. Environmental Protection Agency.
- Pankow, J.F. (1994) An absorption model of gas/particle partitioning of organic compounds in the atmosphere, *Atmos. Environ.*, **28**, 185–188.
- Park, J.S., Jee, N.-Y. and Jeong, J.-W. (2014) Effects of types of ventilation system on indoor particle concentrations in residential buildings, *Indoor Air*, **24**, 629–638.
- Persily, A.K. (2015) Field measurement of ventilation rates, *Indoor Air*, doi:10.1111/ina.12193.
- Persily, A. and Gorfain, J. (2004) *Analysis of Ventilation Data from the U.S. Environmental Protection Agency Building Assessment Survey and Evaluation (BASE) Study*, Report NISTIR 7145, Washington, National Institute of Standards and Technology.
- Persily, A., Musser, A. and Emmerich, S.J. (2010) Modeled infiltration rate distributions for U.S. housing, *Indoor Air*, **20**, 473–485.
- Phillips, L.J. and Moya, J. (2014) Exposure factors resources: contrasting EPA's Exposure Factors Handbook with international sources, *J. Expo. Sci. Environ. Epidemiol.*, **24**, 233–243.
- Polidori, A., Turpin, B., Meng, Q.Y., Lee, J.H., Weisel, C., Morandi, M., Colome, S., Stock, T., Winer, A., Zhang, J., Kwon, J., Alimokhtari, S., Shendell, D., Jones, J., Farrar, C. and Maberti, S. (2006) Fine organic particulate matter dominates indoor-generated PM_{2.5} in RIOPA homes, *J. Expo. Sci. Environ. Epidemiol.*, **16**, 321–331.
- Ramos, T. and Stephens, B. (2014) Tools to improve built environment data collection for microbial ecology investigations, *Build. Environ.*, **81**, 243–257.
- Rehfuess, E., Mehta, S. and Pruss-Ustun, A. (2006) Assessing household solid fuel use: multiple implications for the millennium development goals, *Environ. Health Perspect.*, **114**, 373–378.
- Richardson and Stantec Consulting Ltd (2013) *2013 Canadian Exposure Factors Handbook*, Saskatoon, SK, Canada, Toxicology Centre, University of Saskatchewan, www.usask.ca/toxicology.
- Riley, W.J., McKone, T.E., Lai, A.C.K. and Nazaroff, W.W. (2002) Indoor particulate matter of outdoor origin: importance of size-dependent removal mechanisms, *Environ. Sci. Technol.*, **36**, 200–207.
- Roetzel, A., Tsangrassoulis, A., Dietrich, U. and Busching, S. (2010) A review of occupant control on natural ventilation, *Renew. Sustain. Energy Rev.*, **14**, 1001–1013.
- Santamouris, M., Synnefa, A., Assimakopoulos, M., Livada, I., Pavlou, K., Papaglastra, M., Gaitani, N., Kolokotsa, D. and Assimakopoulos, V. (2008) Experimental investigation of the air flow and indoor carbon dioxide concentration in classrooms with intermittent

- natural ventilation, *Energy Build.*, **40**, 1833–1843.
- Sarnat, S.E., Coull, B.A., Ruiz, P.A., Koutrakis, P. and Suh, H.H. (2006) The influences of ambient particle composition and size on particle infiltration in Los Angeles, CA, Residences, *J. Air Waste Manag. Assoc.*, **56**, 186–196.
- Schneider, A. and Woodcock, C.E. (2008) Compact, dispersed, fragmented, extensive? A comparison of urban growth in twenty-five global cities using remotely sensed data, pattern metrics and census information, *Urban Stud.*, **45**, 659–692.
- Schweizer, C., Edwards, R.D., Bayer-Oglesby, L., Gauderman, W.J., Ilacqua, V., Jantunen, M.J., Lai, H.K., Nieuwenhuijsen, M. and Kunzli, N. (2007) Indoor time-microenvironment-activity patterns in seven regions of Europe, *J. Expo. Sci. Environ. Epidemiol.*, **17**, 170–181.
- See, S.W. and Balasubramanian, R. (2011) Characterization of fine particle emissions from incense burning, *Build. Environ.*, **46**, 1074–1080.
- Seppänen, O.A., Fisk, W.J. and Mendell, M.J. (1999) Association of ventilation rates and CO₂-concentrations with health and other responses in commercial and institutional buildings, *Indoor Air*, **9**, 226–252.
- Seto, K.C. and Fragkias, M. (2005) Quantifying spatiotemporal patterns of urban land-use change in four cities of China with time series landscape metrics, *Landscape Ecol.*, **20**, 871–888.
- Shen, G., Wang, W., Yang, Y., Ding, J., Xue, M., Min, Y., Zhu, C., Shen, H.M., Li, W., Wang, B., Wang, R., Wang, X., Tao, S. and Russell, A.G. (2011) Emissions of PAHs from indoor crop residue burning in a typical rural stove: emission factors, size distributions, and gas-particle partitioning, *Environ. Sci. Technol.*, **45**, 1206–1212.
- Shi, S. and Zhao, B. (2012) Comparison of the predicted concentration of outdoor originated indoor polycyclic aromatic hydrocarbons between a kinetic partition model and a linear instantaneous model for gas-particle partition, *Atmos. Environ.*, **59**, 93–101.
- Shi, S., Chen, C. and Zhao, B. (2015) Air infiltration rate distributions of residences in Beijing, *Build. Environ.*, **92**, 528–537.
- Singer, B.C., Coleman, B.K., Destailhats, H., Hodgson, A.T., Lunden, M.M., Weschler, C.J. and Nazaroff, W.W. (2006) Indoor secondary pollutants from cleaning product and air freshener use in the presence of ozone, *Atmos. Environ.*, **40**, 6696–6710.
- Sippola, M.R. and Nazaroff, W.W. (2002) *Particle Deposition from Turbulent Flow: Review of Published Research and Its Applicability to Ventilation Ducts in Commercial Buildings*, LBNL-51432, Berkeley, Lawrence Berkeley National Laboratory.
- Smith, K.R. (1988) Air pollution: assessing total exposure in developing countries, *Environment*, **30**, 16–35.
- Statistics Canada (2011) General social survey – 2010 Overview of the time use of Canadians, <http://www.statcan.gc.ca/pub/89-647-x/89-647-x2011001-eng.htm>
- Stephens, B. (2015) Building design and operational choices that impact indoor exposures to outdoor particulate matter inside residences, *Sci. Technol. Built Environ.*, **21**, 3–13.
- Stephens, B. and Siegel, J.A. (2012) Comparison of test methods for determining the particle removal efficiency of filters in residential and light-commercial central HVAC systems, *Aerosol Sci. Technol.*, **46**, 504–513.
- Stephens, B. and Siegel, J.A. (2013) Ultrafine particle removal by residential heating, ventilating, and air-conditioning filters, *Indoor Air*, **23**, 488–497.
- Stephens, B., Siegel, J.A. and Novoselac, A. (2011) Operational characteristics of residential and light-commercial air-conditioning systems in a hot and humid climate zone, *Build. Environ.*, **46**, 1972–1983.
- Stephens, B., Azimi, P., El Orch, Z. and Ramos, T. (2013) Ultrafine particle emissions from desktop 3D printers, *Atmos. Environ.*, **79**, 334–339.
- Stewart, J. and Ren, Z. (2003) Prediction of indoor gaseous pollutant dispersion by nesting sub-zones within a multizone model, *Build. Environ.*, **38**, 635–643.
- Stewart, J. and Ren, Z. (2006) COWZ—A subzonal indoor airflow, temperature and contaminant dispersion model, *Build. Environ.*, **41**, 1631–1648.
- Sundell, J., Lindvall, T. and Stenberg, B. (1994) Associations between type of ventilation and air flow rates in office buildings and the risk of SBS-symptoms among occupants, *Environ. Int.*, **2**, 239–251.
- Sundell, J., Levin, H., Nazaroff, W.W., Cain, W.S., Fisk, W.J., Grimsrud, D.T., Gynzelberg, F., Li, Y., Persily, A.K., Pickering, A.C., Samet, J.A., Spengler, J.D., Taylor, S.T. and Weschler, C.J. (2011) Ventilation rates and health: multidisciplinary review of the scientific literature, *Indoor Air*, **21**, 191–204.
- Thatcher, T.L. and Layton, D.W. (1995) Deposition, resuspension, and penetration of particles within a residence, *Atmos. Environ.*, **29**, 1487–1497.
- Thatcher, T.L., Lai, A.C.K., Moreno-Jackson, R., Sextro, R.G. and Nazaroff, W.W. (2002) Effects of room furnishings and air speeds on particle deposition rates indoors, *Atmos. Environ.*, **36**, 1811–1819.
- Thatcher, T.L., Lunden, M.M., Revzan, K.L., Sextro, R.G. and Brown, N.J. (2003) A concentration rebound method for measuring particle penetration and deposition in the indoor environment, *Aerosol Sci. Technol.*, **37**, 847–864.
- Thornburg, J.W., Rodes, C.E., Lawless, P.A., Stevens, C.D. and Williams, R.D. (2004) A pilot study of the influence of residential HAC duty cycle on indoor air quality, *Atmos. Environ.*, **38**, 1567–1577.
- Torkmahalleh, M.A., Goldasteh, I., Zhao, Y., Udochu, N.M., Rossner, A., Hopke, P.K. and Ferro, A.R. (2012) PM_{2.5} and ultrafine particles emitted during heating of commercial cooking oils, *Indoor Air*, **22**, 483–491.
- Turpin, B.J., Weisel, C.P., Morandi, M., Colome, S., Eisenreich, S. and Buckley, B. (2007) Relationships of indoor, outdoor and personal air (RIOPA): part II. Analysis of concentrations of particulate matter species, *Research Report (Health Effects Institute)*, **130**, 79–92.
- United Nations (2008) Principles and recommendations for population and housing censuses, Revision 2, <http://unstats.un.org/unsd/demo-graphic/sources/census/census3.htm>, accessed: 7 June 2014.
- United Nations (2013) Tabulations on households characteristics – data from 2000 and 2010 rounds of censuses, Demographic Yearbook, http://unstats.un.org/unsd/demographic/products/dyb/dyb_Household/dyb_household.htm, accessed: 7 June 2014.
- U.S. Census Bureau (USCB) (2010) Profile of General Population and Housing Characteristics: 2010 Demographic Profile Data. U.S. Census Bureau, accessed: 7 June 2014.
- U.S. Environmental Protection Agency (US EPA) (2011) *Exposure Factors Handbook*, 2011 edn, Washington, DC, National Center for Environmental Assessment; EPA/600/R-09/052F. Available from the National Technical Information Service, Springfield, VA, and online at <http://www.epa.gov/ncea/efh>.
- Vespa, J., Lewis, J.M. and Kreider, R.M. (2013) *America's Families and Living Arrangements: 2012 Population Characteristics*, Washington, DC, U.S. Census Bureau.
- Wainman, T., Zhang, J., Weschler, C.J. and Lioy, P.J. (2000) Ozone and limonene in *Indoor Air*: a source of submicron particle exposure, *Environ. Health Perspect.*, **108**, 1139–1145.
- Wallace, L. (2006) Indoor sources of ultrafine and accumulation mode particles: size distributions, size-resolved concentrations, and source strengths, *Aerosol Sci. Technol.*, **40**, 348–360.
- Wallace, L.A., Emmerich, S.J. and Howard-Reed, C. (2002) Continuous measurements of air change rates in an occupied house for 1 year: the effect of temperature, wind, fans, and windows, *J. Expo.*

- Anal. Environ. Epidemiol.*, **12**, 296–306.
- Wallace, L.A., Emmerich, S.J. and Howard-Reed, C. (2004) Source strengths of ultra-fine particles due to cooking with a gas stove, *Environ. Sci. Technol.*, **38**, 2304–2311.
- Wallace, L., Kindzierski, W., Kearney, J., MacNeill, M., Héroux, M.-È. and Wheeler, A.J. (2013) Fine and ultrafine particle decay rates in multiple homes, *Environ. Sci. Technol.*, **47**, 12929–12937.
- Walton, G.N. and Dols, W.S. (2013) *CON-TAM 2.4 User Guide and Program Documentation*, Gaithersburg, MD, National Institute of Standards and Technology.
- Wang, L. and Chen, Q. (2007) Theoretical and numerical studies of coupling multi-zone and CFD models for building air distribution simulations, *Indoor Air*, **17**, 348–361.
- Wang, H., He, C., Morawska, L., McGarry, P. and Johnson, G. (2012) Ozone-initiated particle formation, particle aging, and precursors in a laser printer, *Environ. Sci. Technol.*, **46**, 704–712.
- Wargocki, P., Sundel, J., Bischof, W., Brun-drett, G., Fanger, P.O., Gyntelberg, F., Hanssen, S.O., Harrison, P., Pickering, A., Seppänen, O. and Wouters, P. (2002) Ventilation and health in non-industrial indoor environments: report from a European Multidisciplinary Scientific Consensus Meeting (EUROVEN), *Indoor Air*, **12**, 113–128.
- Waring, M.S. (2014) Secondary organic aerosols in residences: predicting its fraction of fine particle mass and determinants of formation strength, *Indoor Air*, **24**, 376–389.
- Waring, M.S. and Siegel, J.A. (2008) Particle loading rates for HVAC filters, heat exchangers, and ducts, *Indoor Air*, **18**, 209–224.
- Waring, M.S. and Siegel, J.A. (2010) The influence of HVAC systems on indoor secondary organic aerosol formation, *ASHRAE Trans.*, **116**, 556–571.
- Waring, M.S. and Siegel, J.A. (2013) Indoor secondary organic aerosol formation initiated from reactions between ozone and surface-sorbed D-limonene, *Environ. Sci. Technol.*, **47**, 6341–6348.
- Waring, M.S., Wells, J.R. and Siegel, J.A. (2011) Secondary organic aerosol formation from ozone reactions with single terpenoids and terpenoid mixtures, *Atmos. Environ.*, **45**, 4235–4242.
- Weschler, C.J. (2000) Ozone in indoor environments: concentration and chemistry, *Indoor Air*, **10**, 269–288.
- Weisel, C.P., Zhang, J., Turpin, B.J., Morandi, M.T., Colome, S., Stock, T.H., Spektor, D.M., Korn, L., Winer, A., Alimokhtari, S., Kwon, J., Mohan, K., Harrington, R., Giovanetti, R., Cui, W., Afshar, M., Maberti, S. and Shendell, D. (2005) Relationship of Indoor, Outdoor and Personal Air (RIOPA) study: study design, methods and quality assurance/control results, *J. Expo. Anal. Environ. Epidemiol.*, **15**, 123–137.
- Weschler, C.J. (2006) Ozone's impact on public health: contributions from indoor exposures to ozone and products of ozone-initiated chemistry, *Environ. Health Perspect.*, **114**, 1489–1496.
- Weschler, C.J. (2011) Chemistry in indoor environments: 20 years of research, *Indoor Air*, **21**, 205–218.
- Weschler, C.J. and Nazaroff, W.W. (2008) Semivolatile organic compounds in indoor environments, *Atmos. Environ.*, **42**, 9018–9040.
- Weschler, C.J. and Nazaroff, W.W. (2010) SVOC partitioning between the gas phase and settled dust indoors, *Atmos. Environ.*, **44**, 3609–3620.
- Weschler, C.J. and Shields, H.C. (1999) Indoor ozone/terpene reactions as a source of indoor particles, *Atmos. Environ.*, **33**, 2301–2312.
- Weschler, C.J., Shields, H.C. and Shah, B.M. (1996) Understanding and reducing indoor concentration of submicron particles at a commercial building in southern California, *J. Air Waste Manag. Assoc.*, **46**, 291–299.
- Williams, P.R.D. and Unice, K. (2013) Field study of air exchange rates in northern Highlands of Peru, *Environ. Forensic.*, **14**, 215–229.
- World Health Organization (WHO) (2002) *The Health Effects of Indoor Air Pollution Exposure in Developing Countries*, Geneva, Switzerland, WHO.
- World Health Organization (WHO) (2013) *Review of Evidence on Health Aspects of Air Pollution – REVIHAAP Project: Technical Report*, Geneva, Switzerland, WHO.
- Xiao, J., Shen, Y., Ge, J., Tateishi, R., Tang, C., Liang, Y. and Huang, Z. (2006) Evaluating urban expansion and land use change in Shijiazhuang, China, by using GIS and remote sensing, *Landscape Urban Plann.*, **75**, 69–80.
- Youssefi, S. and Waring, M.S. (2014) Transient secondary organic aerosol formation from limonene ozonolysis in indoor environments: impacts of air exchange rates and initial concentration ratios, *Environ. Sci. Technol.*, **48**, 7899–7908.
- Zaatari, M., Nirlo, E., Jareemit, D., Crain, N., Srebric, J. and Siegel, J. (2014) Ventilation and indoor air quality in retail stores: a critical review (RP-1596), *HVAC & R. Res.*, **20**, 276–294.
- Zhang, Z. and Chen, Q. (2007) Comparison of the Eulerian and Lagrangian methods for predicting particle transport in enclosed spaces, *Atmos. Environ.*, **41**, 5236–5248.
- Zhang, X., Arnot, J.A. and Wania, F. (2014) Model for screening-level assessment of near-field human exposure to neutral organic chemicals released indoors, *Environ. Sci. Technol.*, **48**, 12312–12319.
- Zhao, B. and Guan, P. (2007) Modeling particle dispersion in personalized ventilated room, *Build. Environ.*, **42**, 1099–1109.
- Zhao, B., Yang, C., Yang, X. and Liu, S. (2008) Particle dispersion and deposition in ventilated rooms: testing and evaluation of different Eulerian and Lagrangian models, *Build. Environ.*, **43**, 388–397.
- Zhao, Y., Wang, S., Chen, G., Wang, F., Aunan, K. and Hao, J. (2009) Microenvironmental time-activity patterns in Chongqing, China, *Front. Environ. Sci. Eng. China*, **3**, 200–209.
- Zota, A., Adamkiewicz, G., Levy, J.I. and Spengler, J.D. (2005) Ventilation in public housing: implications for indoor nitrogen dioxide concentrations, *Indoor Air*, **15**, 393–401.
- Zuraimi, M.S., Weschler, C.J., Tham, K.W. and Fadeyi, M.O. (2007) The impact of building recirculation rates on secondary organic aerosols generated by indoor chemistry, *Atmos. Environ.*, **41**, 5213–5223.