

Review

A Scoping Review of the Effects of Ambient Air Quality on Cognitive Frailty

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Abstract: Environmental and public health research has given considerable attention to the impact of air quality on brain health, with systematic reviews being widespread. No literature review has been conducted for cognitive frailty—a multidimensional syndrome combining physical frailty and cognitive impairment and their apparent co-dependence, linked to increased vulnerability and adverse health outcomes, including dementia. Instead, cognitive decline and frailty are implicitly explored through research on air quality and comorbid cognitive and physical decline in elderly populations. A scoping review was conducted to explore the need for a systematic review. Combining the Arksey and O'Malley, and PRISMA-ScR checklist, a scoping review of SCOPUS using 'cogniti*' + 'resilience' + 'air quality' or 'cogniti*' + 'ageing' + 'air quality' resulted in $n = 2503$ articles, screened and reduced using inclusion and exclusion criteria, to $n = 16$ articles. Air quality appears to be a critical risk factor for cognitive decline, even at air quality levels below WHO targets. Moderate long-term ambient air pollution appears linked to increased risk of cognitive frailty, suggesting earlier and more active interventions to protect older people. There are varied effects on cognition across the life course, with both emotional and functional impacts. Effects may be more detrimental to elderly people with existing conditions, including economic and health inequalities. Generalisation of results is limited due to the absence of a dose–response, variations in methods, controlling for comorbid effects, and variance across studies. No literature review has been performed for cognitive frailty, largely due to the fact that it is not presently treated as an explicit outcome. The findings support the need for more research and a more extensive summary of the literature but suggest that there is worsening cognitive function over the life course as a result of increased PM_{2.5} concentrations. Furthermore, air quality appears to be a critical risk factor even at levels below World Health Organisation targets.

Keywords: cognitive function; ambient air quality; air pollution and brain health; older adults; cognitive frailty



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1. Introduction

Following [1,2], there is growing evidence that exposure to high levels of air pollutants across the life course, particularly in old age, is detrimental to brain health and increases the risk of dementia syndrome and cognitive decline [3]. In the elderly, this impact ranges from

global cognition and neurodegenerative disorders to neuroinflammation and cognitive frailty [4–6].

Our scoping review focused on *cognitive frailty*—a new term for describing the simultaneous presence of physical frailty and cognitive impairment without dementia [7,8].

While the definitions and underlying mechanisms of cognitive frailty are still debated, older adults with both frailty and cognitive impairment are at higher risk of adverse health outcomes, including death, disability, hospitalisation, and incident dementia, than those with either condition alone [9]. In some cases, the progression of cognitive frailty can be impeded or even reversed. Understanding modifiable risk factors could lead to interventions to lower the risk of progression to dementia, neurodegenerative disorders, and other mental health consequences [10].

One such modifiable risk factor for cognitive frailty is air pollution. Ambient air pollution is the largest environmental risk to public health, contributing to ~4.2 million premature global deaths each year [11], with significant direct and indirect economic costs. In ref. [12], air pollution is measured through exposure to gases such as nitrogen oxides (NO_x), sulphur dioxide (SO₂), ozone (O₃), and particulate matter (PM). In ref. [13], of particular concern for brain health, including cognitive frailty, is fine particulate matter (particulate matter with an aerodynamic diameter of ≤2.5 μm, PM_{2.5}). The World Health Organisation (WHO) Air Quality Guidelines (AQGs) were updated in 2021, reflecting evidence of adverse health impacts at low exposure levels (Table 1). In ref. [14], as shown, the new 2021 update has a ‘Guideline’ figure for PM_{2.5} of 5 μm⁻³, which is the current lowest level at which adverse health effects will occur due to exposure, and four interim targets. The interim targets are designed to facilitate stepwise improvements in air quality and population health benefits. For PM_{2.5}, each change from the guideline level through to interim target 1 would see an increase in mortality of 4%, 8%, 16% and 25%.

Table 1. Changes in the World Health Organisation’s Air Quality Guidelines.

Pollutant	Averaging Time	WHO 2005 Guideline	WHO 2021 Update				Guideline
			Interim Targets				
			1	2	3	4	
Fine Particles, PM _{2.5}	Daily (24 h) mean (μgm ⁻³)	10	35	25	15	10	5

Various proposed direct and indirect causal mechanisms have been reported on how air pollution impacts health, be it through the respiratory system, the cardiovascular system, the central nervous system, the gastrointestinal system, or epigenetic changes to DNA. In refs. [15–17], in terms of the brain, air pollution impacts health “via key biomarkers, inflammatory processes (neuro and vascular) and cardiopulmonary disease; and there is growing research showing that chronic exposure can deteriorate the protective barriers of the brain” [18] (p. 2). For example, ultrafine particles (with a diameter less than 0.1 μm, PM_{0.1}) can potentially enter the brain via the olfactory bulb, resulting in neuroinflammation and cognitive impairment [19–22].

In terms of cognitive frailty, a recent nationwide prospective cohort study found a positive association between long-term PM_{2.5} exposure (defined as a 10 μgm⁻³ increment of PM_{2.5}) and a 5.0% increase in the incidence risk of frailty [5]. The estimated disease burden indicated that implementing clean air actions may reduce functional frailty and substantially offset the burden of population ageing worldwide. Evidence is also emerging that greater exposure to airborne pollutants, such as NO_x, is associated with a higher probability of being in both the pre-frail and frail categories (odds ratio (OR) 1.003; 95% CI, 1.001–1.004) [23]. Similarly, the increase in the exposure to PM_{2.5–10} was associated with a higher probability of being pre-frail and frail (OR 1.014; 95% CI [1.001–1.036]), such as the increment in the exposure to PM_{2.5} that was associated with a higher probability of

being frail (OR 1.018; 95%, CI 1.001–1.037) [23]. PM_{2.5} exposure in older adults has also been linked to an increase in their depressive and anxiety symptoms [24].

While such studies have begun to identify key linkages, the research on the relationship between long-term exposure to particulate matter and cognitive frailty and functional disability in the elderly is limited, with no current summary of the literature. This gap in the literature is largely due to the fact that cognitive frailty is not presently treated as an explicit outcome; instead, it is implicitly explored through research on air quality and comorbid cognitive and physical decline in elderly populations, hence the focus of our study. Combining ref. [1] and the *PRISMA-ScR checklist* [2], we piloted a scoping review to (1) investigate the effects of poor air quality on cognitive frailty in older populations and (2) identify gaps in the evidence surrounding this. Following [2], the purpose of our scoping review was to determine if there is value in undertaking a more rigorous and systematic review of the literature and to identify which measures of air quality and cognitive frailty have been used in studies to date, with a view to, first, standardising reported measures for the future, and second, aiding in the planning and commissioning of future research [2] (p. 467).

2. Materials and Methods

2.1. Scoping Review Approach

This scoping review is part of a larger set of three separate scoping reviews that explore the impact of environmental factors on cognitive frailty. For our protocol, we followed Arksey and colleagues [1]. Research questions were derived through their six stages using discussion and searches. Our rationale for using Arksey and colleagues [1]—which Tricco and colleagues [2] also use to construct the *PRISMA-ScR checklist*—is because it is useful for consultation and for policy makers, practitioners and service users whilst adopting transparent methods and global good practice standards for data analysis and presentation [25]. To organise and write up our study according to current practice, we used the *PRISMA-ScR checklist* [2].

2.2. Protocol

The literature database SCOPUS was searched to identify relevant observational and longitudinal studies. The terms ‘cogniti*’ + ‘resilience’ + ‘air quality’ or ‘cogniti*’ + ‘ageing’ + ‘air quality’ were used; and, though aware of the limitations this brings, only results in English were included. Following the initial search, all duplicates were removed, and the titles were screened against inclusion and exclusion criteria. Book chapters, reviews and commentaries were excluded.

Titles/abstracts were screened by two of the lead authors of the study, and their eligibility for the study, based upon the rigorousness of their ambient air quality data, was screened by a lead author and the air quality expert on the team. Titles had to clearly indicate that the study explored the impact of air quality on cognitive aspects of frailty (*excluding* dementia or, specifically, mild cognitive impairment) and that the participants were from an older age group, or a specific older age range. The abstracts of the remaining studies were then screened to identify those that specifically addressed cognition as a part of frailty in older persons because the aetiology of cognitive frailty may be different and is potentially reversible. This stage was important to exclude studies relating exclusively to physical frailty which were the majority.

To manage the findings and limit the range of variables involved in the association between air quality and cognition, the *exclusion criteria* were then amended to exclude all studies relating to indoor air quality or not considering PM_{2.5}, as this aspect of air pollution has the most deleterious health effects and highest risk of negative cognitive impacts [3]. All of the included papers utilised some form of modelled air pollution mapping, validated by national, government-run field monitoring networks, such as the United States Environmental Protection Agency and Department for Food, Environment and Rural Affairs.

An excel spreadsheet was used for data extraction with headers that reflected key aspects of the findings in order to compare and contrast the ways that the studies were conducted, and data reported such as population outcomes. Demographic information was identified (the age range of included participants, with gender and level of education attainment and social or financial status of the participants). The country of data collection, details on the methodology, sample size, and the methods of assessment of both air quality and cognition were recorded. Study limitations and comments were also recorded by the author (CB) and the data extraction sheets were then shared with the wider transdisciplinary group of co-authors for the purpose of summarising and synthesising the data and prioritising the discussion.

Data analysis occurred following data extraction, with cognitive assessment and air quality measures compared within the selected papers. This was to identify how varied methods were used for the older adult cognitive frailty assessment and how the levels of pollutants were examined and reported. Data analysis included the assessment of air quality data utilised, and how this compared to WHO and national air quality guidelines and the inclusion of potential confounding variables used in analyses.

3. Results

3.1. Scoping Review

There were 16 papers selected for inclusion that represented the international literature from both OECD countries (United States, Taiwan, China, and Scotland) and upper-to-middle-income countries including Mexico and Chile. Figure 1 below presents the process of article selection according to the *PRISMA-ScR checklist* [2]. It demonstrates the selection of articles according to refined inclusion and exclusion criteria.

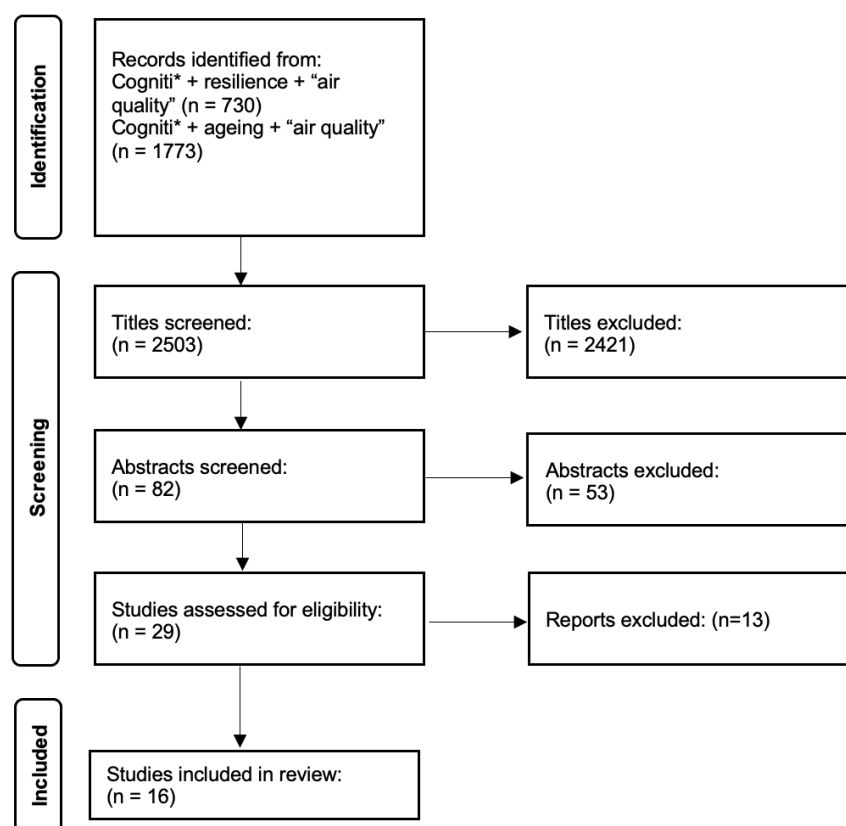


Figure 1. PRISMA-ScR checklist. Deselection following data extraction of (1) indoor air quality (2) data that used atmospheric readings and (3) where the data were not presented in such a way that the mean readings could be compared. All but one paper are quantitative studies and 11 are large cohort analyses of population data.

All but four of the studies included gender-balanced participant groups, with three of these being female only. All but one study was quantitative, with the qualitative study used to study the effects of cardiorespiratory exercise on cognition in older women exposed to air pollution. The remaining studies included those with large cohorts reflecting a demographic of older people ages ranging from 50 to 98. The demographic factors associated with education level and household wealth were also extracted, based on the known shielding effect of higher levels of education and wealth associated with cognitive resilience. As presented in Table 2, the selected papers demonstrate inconsistencies in the reporting of wealth and education but were all included to identify the association between cognition in older adult populations in relation to poor air quality and the risks associated with declining cognition.

3.2. Variations in Air Quality Assessment

As previously highlighted, the 16 papers selected for inclusion represented international literature from The Organization for Economic Cooperation and Development (OECD) referring to wealthier countries contrasting with low to middle income countries (LMIC) countries. PM_{2.5} data utilised in the studies were often modelled through a variety of methods, although the most commonly used was a version of kriging. Seven papers did not use fixed monitoring station data as a basis of their modelling but most of these did use satellite or other monitoring methods. Most but not all studies used static monitoring sources to validate model performance, and this showed strong correlations between models and monitoring stations. Averaging periods for PM_{2.5} exposure for the majority of studies were annual mean concentration, although some utilised 24 h mean or longer exposure periods. Further details are presented in Table 3.

3.3. Overall Results

There is evidence that living in areas with the highest quintiles of cumulative PM_{2.5} was associated with a risk of developing poor cognitive function [26] and that higher levels of exposure to ambient PM_{2.5} are associated with worse cognitive decline [27]. Long-term exposures to PM_{2.5} and NO_x, estimated from participants' addresses, were directly associated with lower global cognitive performance [28]. The associations between particulate air pollution and cognitive function, as well as cognitive decline over time, were also confirmed [29]. Ref. [30] shows that 11-year exposure to elevated PM_{2.5} levels was associated with poor global cognition over four-year follow-up; and, over time, the likelihood of impact from long-term exposure was an increased incidence of 46% [31]. In addition, the Lothian Birth Cohort analysis [6], reported higher air pollution modelled from 1935 (when participants would have been in utero), was associated with a worse change in IQ from age 11–70 years, suggesting that a life course paradigm is essential for understanding cognitive decline in relation to air pollution.

Some studies differentiated the risks to different groups, for example, PM_{2.5} was unrelated to cognition among those with 13 or more years of education [32]. Similarly, exposure to PM_{2.5} was associated with other risk factors in older people, for example, cognitive function appeared to be reduced in those who smoked [32]. For example, ref. [33] identified a higher risk of an effect between PM_{2.5} and the incidence of disability in men, smokers, and participants with cognitive impairment. Ref. [34] recognised that non-Hispanic participants had less cognitive change when exposed to a higher PM_{2.5} concentration; they also lacked the genetic factor (genotype E4 (APOE-ε4) allele) that accelerated the rate of decline. Ref. [35] also observed an association between air pollution with cognitive decline mediated via air pollution impacting the sleep cycle which resulted in brain structure changes. Ref. [31] observed a notable association between cumulative PM_{2.5} and worse cognitive function being more pronounced in participants whose diets scored lower on the plant-based diet index.

Table 2. Selected articles related to cognition, older adults and air pollution.

Authors	Title	Published	Country	OECD /UMIC	Maximum Cohort Number	Gender Male/Female (in %)	Reported Level of Education	Wealth
Ailshire and Walseman [32]	Education differences in the adverse impact of PM _{2.5} on incident cognitive impairment among U.S. older adults	2021	United States	OECD	5636	46/54	59%, 12 years of education	Majority high paid workers
Ailshire and Crimmins [36]	Fine particulate matter air pollution and cognitive function among older US adults	2014	United States	OECD	13,996	44/56	Mean 14 years of education	No data
Chen et al. [30]	Long-Term exposure to air pollutants and cognitive function in Taiwanese community-dwelling older adults: A four-year cohort study	2020	Taiwan	OECD	605 (305 at follow up)	46/54	Mean 14 years of education	No data
Kulick et al. [34]	Long-term exposure to ambient air pollution, APOE-ε4 status, and cognitive decline in a cohort of older adults in northern Manhattan	2020	United States	OECD	4821	32/68	Mean 9 years of education	No data
Lo et al. [35]	Air pollution associated with cognitive decline by the mediating effects of sleep cycle disruption and changes in brain structure in adults	2022	Taiwan	OECD	4866	69/31	Mean 10 years of education	No data
Lv et al. [5]	Long-term exposure to PM _{2.5} and incidence of disability in activities of daily living among oldest old	2023	China	OECD	15,453	44/56	67% illiterate 33% literate	No data
Molina-Sotomayor et al. [37]	Effects of cardiorespiratory exercise on cognition in older women exposed to air pollution	2019	Chile	UMIC	181	All women	No data	No data
Petkus et al. [28]	Outdoor air pollution exposure and inter-relation of global cognitive performance and emotional distress in older women	2021	United States	OECD	6118	All women	71% higher educated	Assumed wealthier

Table 2. Cont.

Authors	Title	Published	Country	OECD /UMIC	Maximum Cohort Number	Gender Male/Female (in %)	Reported Level of Education	Wealth
Qiu et al. [38]	Associations between air pollution and psychiatric symptoms in the Normative Aging Study	2022	United States	OECD	570	All men (veterans)	13% higher educated	No data
Russ et al. [6]	Life course air pollution exposure and cognitive decline: Modelled historical air pollution data and the Lothian birth cohort 1936	2021	Scotland UK	OECD	572	53/47	28% higher educated	No data
Salinas-Rodríguez et al. [39]	Exposure to ambient PM _{2.5} concentrations and cognitive function among older Mexican adults	2018	Mexico	UMIC	7986	46/54	No data	No data
Tonne et al. [29]	Traffic-related air pollution in relation to cognitive function in older adults	2014	UK	OECD	3414	65/35	39% higher educated	41% high paid
Wang et al. [40]	Fine particulate matter and poor cognitive function among Chinese older adults: Evidence from a community-based, 12-year prospective cohort study	2020	China		13,324	48/52	59% illiterate	No data
Weuve et al. [27]	Exposure to particulate air pollution and cognitive decline in older women	2012	United States	OECD	19,409	All women	100% graduate nurses	No data
Yao et al. [26]	The effect of China's Clean Air Act on cognitive function in older adults: a population-based, quasi-experimental study	2022	China	OECD	2812	50/50	No data	No data
Zhu et al. [31]	Interaction between plant-based dietary pattern and air pollution on cognitive function: a prospective cohort analysis of older adults	2022	China	OECD	6525	49/51	48% formal education	35% 'financially independent'

Table 3. Details of monitoring and modelling methods utilised as well as validation and pollution averaging periods.

Authors	Publication	Country	Monitoring Method	Modelling Method	Validation Method	Averaging Period (s)
Ailshire and Walseman [32]	2021	United States	United States Environmental Protection Agency (USEPA) monitoring stations	Gridded 12 km model	None stated—utilising USEAP model	Annual average concentration
Ailshire and Crimmins [36]	2014	United States	USEPA monitoring stations	Inverse distance weighting of monitoring stations to participant census tract	None stated	24 h mean
Chen et al. [30]	2020	Taiwan	Taiwan Environmental Protection Administration monitoring stations	Bayesian maximum entropy	None stated	Annual mean concentration
Kulick et al. [34]	2020	United States	USEPA monitoring stations	Universal kriging	None stated	Annual Average
Lo et al. [35]	2022	Taiwan	Taiwan Environmental Protection Administration monitoring stations	Hybrid kriging-land-use regression	None stated	Daily concentrations aggregate to annual average
Lv et al. [5]	2023	China	Satellite, simulation and monitoring stations utilised to estimate resident exposure.	Not specified	Out-of-sample cross-validation with monitors	Annual concentration over 4 time points (2002, 2005, 2008, 2011)
Molina-Sotomayor et al. [37]	2019	Chile	Not specified	Not specified	Not specified	Annual average (2012, 2013, 2014)
Petkus et al. [28]	2021	United States	USEPA monitoring stations	Regionalised universal kriging	None stated	Annual mean concentration
Qiu et al. [38]	2022	United States	USEPA monitoring stations, satellite-derived measurements	Machine-learning algorithms using geographically weighted regression	Cross-validated to monitored values	Averages of 1 week, 4 weeks, 8 weeks and one year prior to the visit
Russ et al. [6]	2021	Scotland UK	N/A	EMEP4UK atmospheric chemistry transport model	Validated against UK monitoring stations	Annual mean PM2.5 concentration (1935, 1950, 1970, 1980, 1990, 2001)

Table 3. Cont.

Authors	Publication	Country	Monitoring Method	Modelling Method	Validation Method	Averaging Period (s)
Salinas-Rodríguez et al. [39]	2018	Mexico	N/A	Aerosol optical depth information retrieved from remote sensing data. Combination of observations from the Moderate Resolution Imaging Spectroradiometer and Multiangle Imaging Spectroradiometer instruments on the Terra satellite (NASA satellites), and simulations with the Goddard Earth Observing System (GEOS) chemical transport model.	None stated	Annual average
Tonne et al. [29]	2014	UK	N/A	KCLurban and CMAQ-Urban models	Validated against monitoring stations	1-, 3- and 5-year average
Wang et al. [40]	2020	China	N/A	Utilised aerosol optical depth data with a land-use regression model.	Global ground-based stations and geographical weighted regression method.	Averaged for study period as well as 3 years prior to recruitment
Weuve et al. [27]	2012	United States	USEPA monitoring data	Geographic information system-based spatiotemporal smoothing	Not stated	1 month and 7–14 year exposures
Yao et al. [26]	2022	China	N/A	Atmospheric chemical transport model.	Not stated	Annual average concentration
Zhu et al. [31]	2022	China	Remote sensing, Multiangle Imaging SpectroRadiometer and Sea-viewing Wide Field-of-view Sensor satellite instrument, vertical profiles derived from the GEOS-Chem chemical transport model	Utilised monitoring data previously described	Ground-based observations using geographically weighted regression	Annual mean from 1998 to 2014

The presence of different covariates is important to note, given the number of studies that seek to adjust for community social characteristics and individual social, demographic, economic, and health characteristics. In some studies, area-level household income and house value modified the associations between ozone exposure and cognitive effects, with residents of lower-income or -house-value neighbourhoods having an increased response to ozone [38]. However, in [28], the emotional distress associated with (in particular) NO_x was not confounded by socio-economic factors (age, geographical region, race/ethnicity, education, and income), lifestyle (smoking, alcohol, and physical activity), employment status or clinical characteristics [28].

The level of disability caused in more vulnerable populations from $\text{PM}_{2.5}$ exposure means that the threshold for significant risk is important. Wang et al. (2020) identified that each $10 \mu\text{g m}^{-3}$ increase in $\text{PM}_{2.5}$ was associated with a 5.1% increased risk in poor cognitive function. Subgroup analyses suggested stronger associations between $\text{PM}_{2.5}$ and poor cognitive impairment in men than women [41]. Ref. [38] also notes a positive association between short to medium term (average one and four weeks) ambient exposure to gaseous pollutants (ozone and NO_x) and elevated psychiatric symptom intensity. One study provided limited population-based evidence for the oldest subjects and detected a threshold of $33 \mu\text{g m}^{-3}$. Below this level, there may be a lower risk of disability in the activities of daily living [33]. The operative thresholds of functional disability enable the consideration of consequent needs arising from cognitive impairment.

Mean $\text{PM}_{2.5}$ relative to the WHO target air quality guidelines is shown in Figure 2 below. All but one of the sixteen studies [38] showed a detrimental impact of $\text{PM}_{2.5}$ on cognitive function. Levels of $\text{PM}_{2.5}$ that negatively influenced cognition varied from the lowest detrimental concentration being $7.9 \mu\text{g m}^{-3}$, below WHO Interim Target 4 ($10 \mu\text{g m}^{-3}$), up to $50.3 \mu\text{g m}^{-3}$.

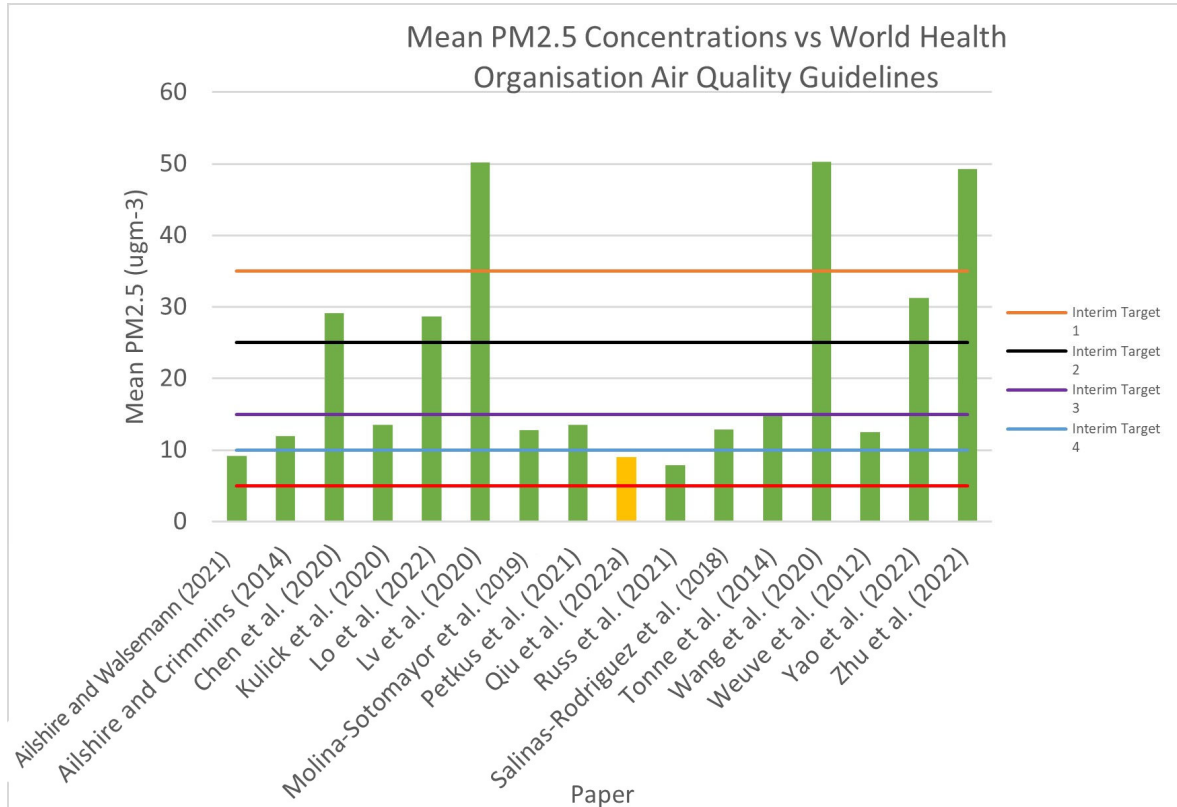


Figure 2. Mean $\text{PM}_{2.5}$ concentrations reported in the 16 studies mapped against WHO AQGs. Green bars indicate a detrimental relationship between $\text{PM}_{2.5}$ and cognition variables. Amber bars indicate no relationship was found [5,6,26–32,34–40].

3.4. Variations in Cognitive Assessments

Some studies focussed on memory assessment as a means of measuring cognition, as in [32] who used 10-word immediate and delayed recall and serial seven subtraction (attention and processing speed, working memory and counting backwards) to assess incident cognitive decline. Similar methods for immediate and delayed 10-noun free recall tests to measure memory, included a serial seven subtraction test to measure working memory, a backwards-counting test to measure processing speed, recall the date and name of the president and vice president to measure orientation, and an object-naming test to measure knowledge and language [36]. The focus on working memory being adversely affected by exposure to PM_{2.5} suggests that older people's long-term memory and recollection of previous experiences may also be affected by other ageing phenomena or accelerated by the effect of pollution on the cardiovascular and other systems [32,36].

Functionally oriented tests perhaps provide a more reliable measure of change over time that can be attributed to personal characteristics such as intelligence and individual cognitive resilience. Ref. [30], for example, used the Montreal Cognitive Assessment-Taiwanese version (MoCA-T, [42]) to test domain-specific cognition, e.g., memory, executive function, category/semantic fluency, and attention. They also used the Wechsler Memory Scale, 3rd edition, to assess attention performance and verbal fluency tests. Ref. [29] used the Alice Heim 4-I test (see NSHD—Age 15—The Alice Heim Group Ability Test (AH4)—CLOSER) for global intelligence—comprised of a series of 65 verbal and mathematical reasons items of increasingly difficulty—to measure the impact of air quality on inductive reasoning and the ability to identify patterns and infer principles and rules. They also used a 20-word recall test for short-term verbal memory; and two measures of verbal fluency to measure cognitive function and decline over time in cognition. These studies, along with [6], also sought to identify protective factors, as in the case of educational level and diet, which may provide resilience to cognitive frailty. Ref. [6] also used Lothian Birth Cohort data collected from the Moray House Test in 1947 (for subjects born in 1936) and considered whether risk is accumulated through long-term exposure or whether there are sensitive or critical periods in the life course.

In contrast, ref. [33] used the Katz Index of Independence in Activities of Daily Living (alz.org) with the oldest elderly Chinese subjects in their study to assess functional changes in relation to PM_{2.5} and showed a decline in relation to exposure. The rationale for functional assessment to measure cognition is well established and aligns with the International Classification of Functioning, Disability and Health (ICF) (who.int), which is an assessment of health and health-related domains and the disability of an individual that occurs in a context. The ICF also includes a list of environmental factors that impact the ability to perform tasks, and which are meaningful to individuals, such as the presence of support/carer/family in the environment.

Several studies, e.g., [32,39], used specific dementia assessments to identify those with normal cognition, such as the Data_Description_Langa_Weir_Classifications2020.pdf (umich.edu). Commonly, the Mini-Mental State Exam, CASI (C-2.0), is used to measure orientation, memory registration, concentration, delayed recall memory, language and construction in studies that are investigating the impact of air quality on these factors. CASI: attention, orientation, short-term memory, long-term memory, language abilities, mental manipulation, verbal fluency, abstract thinking, drawing (see mini-mental state examination (MMSE)—Oxford Medical Education). These functional tests were used in six of the 16 studies and are appropriate for the identification of working memory, also used by [28,37]. These latter studies aimed to identify the association between long-term ambient air pollution exposure, emotional distress and cognitive decline in older adulthood as a precursor to a psychiatric condition.

3.5. Limitations of the Selected Studies

The first limitation concerns assessing or controlling for covariate factors. The 16 studies selected for this scoping review all reported mean levels of particulates. Potential studies

that did not report a mean were excluded. Doing so allowed us to compare pollution levels and their impact on cognitive frailty across studies. These other studies aside, when looking at the 16 studies in this scoping review, while the methods of analysis for cohort data were deemed reliable, they did not control for the presence of multiple covariate factors, e.g., pre-existing conditions, the effect of long-term poverty, low paid work, and an assumed lack of resilience to other environmental factors, all of which negatively impact people in older age groups [43]. Future studies need to take these covariates into consideration.

The second issue is measurement error for air pollution estimates, e.g., PM_{2.5} and NO_x exposures [28]. Similarly, ref. [6] highlights that results have large degrees of uncertainties, considering the various methodologies used to produce the different air pollution concentration estimates. Furthermore, all but one study [37] compared pollution concentrations to WHO Air Quality Guidelines and interim targets and/or national guideline concentrations. This is an area that would be beneficial to be included in future studies to allow for greater comparison against WHO standards and help map out reduction pathways for local authorities to improve air quality and population exposure.

4. Discussion

4.1. Limitations of Scoping Review and Gaps in the Literature

This scoping review collated evidence on the impact that ambient air quality has on various measures of cognitive outcomes in older adult populations, known more recently as multi-dimensional cognitive frailty. The value of this review was that, presently, no such summary of the literature has been conducted. Our goal was to (1) investigate what the literature has to say on the effects of poor air quality on cognitive frailty in older populations and (2) to see if we could identify gaps in the evidence surrounding this. Following [2], we also sought, more widely, to determine if there is value in undertaking a more rigorous and systematic review of this literature and to identify which measures of air quality and cognitive frailty have been used in studies to date, with a view to, first, standardising reported measures for the future, and second, aiding in the planning and commissioning of future research.

As concerns our primary goal, we can conclude that, yes, there is a need for a more systematic review of the literature and that there are a number of limitations and gaps in the literature that we identified which can inform future research. First, given that our study was a scoping review, we did not provide an assessment of quality for the 16 studies. The low number of studies makes judgement of the quality of the body of evidence available difficult. Any future in-depth summary of the literature would benefit from such an assessment. Second, cognitive frailty is a new term, and has only more recently been used as a defined and explicit condition. While the studies we reviewed measure different aspects of cognitive frailty in the elderly, it is essential, going forward, that future research address this gap by making cognitive frailty an explicit topic of study. Third, as mentioned earlier, another gap in the literature concerns differences across studies in the degree to which they assessed or controlled for covariates, making it difficult to determine the overall impact of air quality on frailty. Finally, as with most literature reviews, there is the limitation of making generalisations due to differences in how cognitive outcomes are measured across studies.

We also acknowledge that, given our focus on SCOPUS, certain studies may have been missed due to not being part of the database or key-worded differently. Also, given that the definition of cognitive frailty is somewhat contested or not known by researchers in the broader field of air quality and brain health, there is a chance authors of certain studies could have used other terms.

We also recognised that, in terms of another gap in the literature, there are no studies on this topic for low-to-middle-income countries, resulting in a bias towards upper middle income and OECD countries. Another is the absence of rigorous socio-demographic data or in-depth reporting sufficient to engage in a summary review of the role socioeconomic deprivation or inequalities play in the impact air pollution has on the elderly across different

countries. Similarly, the potential of higher education and wealth to ameliorate the risks of frailty and cognitive impairment were inaccessible due to the failure to systematically report demographic details. This is important because the differentiated health effects from poor air quality are experienced at a range of scales, with demography, density and health geography all playing a part [44–46]. In most instances, the oldest and most disadvantaged within a community experience the worst pollution levels, while also often lacking the ability to negate the detrimental effects of pollution exposure [47,48]. In addition to air quality, the socio-ecological systems that people live and spend time in, either for work or leisure are important factors when considering health effects [44–46]. The effects of air quality are also exacerbated by dementia and other health risks [49].

4.2. Standardising Measures of Air Quality and Cognitive Frailty

Following [2], we also concluded that our findings warrant a more rigorous and systematic review of the literature. In terms of standardisation of measures, we established that measures of PM_{2.5} are the most useful, particularly when reported as means and with the impact of covariates on cognitive frailty outcomes taken into consideration. Ultrafine particles are an important pollutant but there are no studies utilising UFPs in terms of cognition or frailty, likely due to difficulties in accurate and reliable UFP measurement techniques as highlighted by Marval and Tronville’s review [50]. In terms of measuring cognitive frailty, in addition to the need to make this outcome more explicitly measured and rigorous, we identified several cognitive measures used, including tests for immediate and delayed recall, episodic memory, and cognitive functionality, including the Montreal Cognitive Assessment-Taiwanese version (MoCA-T, [36]); the Wechsler Memory Scale, 3rd edition; and the Alice Heim Group Ability Test (AH4). Finally, to measure independence, the Katz Index of Independence in Activities of Daily Living was used. Functionality was particularly important because it provides a more reliable measure of cognitive change over time.

4.3. Thinking about Policy and Practice

The increasing number of older adults across the world represents a demographic megatrend that is transforming economies and societies [51]. While all countries are experiencing an ageing of their populations, their respective health trends manifest along different trajectories, mostly as a function of differences in environmental and socio-demographic factors, as in the case of air pollution [40,52]. The incidence of frailty in older people is increasing too and measures that reduce the risk of disease and promote the maintenance of function, confidence and engagement are important to support healthy ageing and ease the pressure on healthcare systems [53].

In this context, this scoping review aimed to identify the association between air pollution and cognitive frailty in older people without dementia. The incidence of cognitive frailty (1–12% of the adult < 65 population) is now being recognised as a cause of subsequent disability. The identification of cognitive impairment as an intersectional feature of frailty provides an opportunity for secondary prevention: alleviating cognitive impairment in older adults improves their functional health [8]. In addition to programmes to ameliorate cognitive frailty from either a cognitive or physical approach, we must also address underlying mediatory factors such as air pollution that live outside of the control of the individual.

The intersection between climate change and population ageing has only begun to receive attention, despite being considered a critical global issue [54]. Older populations are particularly vulnerable, often with pre-existing morbidities that are sensitive to the consequences of climate change, such as extreme heat and air pollution [55]. Green planning initiatives such as improving air quality may reduce the future population’s burden of poor cognitive function, especially in areas where ambient air pollution is identified as a risk factor for older adults. However, current policy initiatives are associated with the reduction of the ‘impact’ from pollution and not the reduction of air pollution per se. For example,

the UK government has announced that councils in England have been granted almost £11 million in funding for projects that will improve air quality [56]. This annual 'Air Quality Grant' helps councils develop and implement measures to benefit schools, businesses, and communities by reducing the impact of air pollution on people's health. Older adults are not a homogenous group and councils will need to decide on the targeting of poorer communities (where air pollution is often the highest) or on more general interventions such as clean air zones [57]. In addition, a systematic review published in *The Lancet* has suggested that low-emission air zones can reduce air-pollution-related health outcomes, and that ongoing evaluation of these interventions may help explore their role in reducing cognitive frailty in the hope that high-risk and vulnerable groups benefit alongside the general population [58].

4.4. *The Differential Impact of Air Quality, Even When Guidelines Are Met*

Differential levels of air pollution and their effects on cognition and frailty can be seen in the selected literature. This includes studies demonstrating that even low levels of air pollution ($<10 \mu\text{g m}^{-3}$) can have a harmful health effect, particularly over long periods. This effect is also often below the WHO Interim Target 3 concentration of $15 \mu\text{g m}^{-3}$, which was highlighted as being close to the mean $\text{PM}_{2.5}$ concentration reported in long-term exposure studies [58]. The WHO AQGs and interim targets have been designed to provide a framework for authorities to work towards and through, which, coupled with our scoping review, highlights the need to increase the number of long-term exposure studies determining the life course impact of even low levels of pollution on cognition [59]. However, with the exception of one paper, none of the 16 articles reviewed here included reference to WHO or national guidelines for pollution concentrations [37].

4.5. *Thinking about Co-Benefits*

The impacts of air pollution and climate change are often understood as global and considered to occur over a timescale of decades. Viewing these impacts from a co-benefits lens allows them to be considered from a local, short-term perspective which is often more politically relevant. These co-benefits, which have both health and economic advantages, include not only mitigating against or even reversing cognitive frailty, but also, across the life course, reducing premature mortality, sickness days, healthcare expenditures, and poor indoor air quality [60]. Many older adults spend a considerable amount of their time indoors—a topic not addressed by this scoping review, but of importance to any wider summary of the literature. For example, a longitudinal study found that both indoor and ambient air pollution had adverse effects on the mental health of the elderly, particularly with regard to depressive symptoms [61]. It is well established that depression is associated with cognitive impairment and so addressing air pollution may be one way in which reversible elements of cognitive frailty can have other positive brain health effects.

5. Conclusions

Environmental and public health research has given considerable attention to the impact of air quality on brain health, with systematic reviews being widespread. No summary of the literature has been conducted for cognitive frailty—a multidimensional syndrome combining frailty and cognitive impairment, linked to increased vulnerability and adverse health outcomes, including dementia. This gap in the literature is largely due to the fact that cognitive frailty is not presently treated as an explicit outcome; instead, it is implicitly explored through research on air quality and its impact on comorbid cognitive and physical decline in elderly populations, hence the focus of our scoping review.

Combining ref. [1] and the PRISMA-ScR checklist [2], a scoping review of SCOPUS using 'cogniti*' + 'resilience' + 'air quality' or 'cogniti*' + 'ageing' + 'air quality' resulted in $n = 2503$ articles, screened and reduced using inclusion and exclusion criteria, for a total of $n = 16$ articles.

We identified several insights across the literature, albeit tentatively, given the small number of studies. For example, we found evidence of an association between higher PM_{2.5} levels and worse cognitive function and rate of decline in cognitive function over the life course, although certain population factors influenced the size of the effect. We also found that air quality appears to be a critical risk factor even at air quality levels below World Health Organisation targets. There are also varied effects on cognition across the life course, with both emotional and functional impacts. Effects may be more detrimental to elderly people with existing conditions, including economic and health inequalities. Finally, in terms of standardising measures, we identified PM_{2.5} averages as being central, as well as several cognitive measures, with cognitive function being particularly reliable, given the potential reversibility of cognitive frailty.

In terms of the value of these insights for policy and practice, interventions aimed at improving air quality and minimising the impact of exposure to air pollution for those with this particular cognitive impairment could actively address the disability associated with increased frailty and significant additional healthcare utilisation. The benefits of cleaner air are particularly great for older, and more vulnerable elderly groups; but, systemic improvements to air quality will benefit society as a whole and further prevent the lifespan effects of air pollution on neonates, children, and vulnerable adults. The suggested health risk at even low levels of air pollution on cognition needs to be the subject of a major policy initiative for those with recognised vulnerability due to age or pre-existing morbidity. The estimated disease burden indicates that implementing clean air actions may prevent frailty and substantially offset the burden of population ageing worldwide.

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