Long-term air pollution exposure and cardio-respiratory mortality: a review

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Abstract

Current day concentrations of ambient air pollution have been associated with a range of adverse health effects, particularly mortality and morbidity due to cardiovascular and respiratory diseases. In this review, we summarize the evidence from epidemiological studies on long-term exposure to fine and coarse particles, nitrogen dioxide (NO2) and elemental carbon on mortality from all-causes, cardiovascular disease and respiratory disease. We also summarize the findings on potentially susceptible subgroups across studies. We identified studies through a search in the databases Medline and Scopus and previous reviews until January 2013 and performed a meta-analysis if more than five studies were available for the same exposure metric.

There is a significant number of new studies on long-term air pollution exposure, covering a wider geographic area, including Asia. These recent studies support associations found in previous cohort studies on PM2.5. The pooled effect estimate expressed as excess risk per 10 μg/m3 increase in PM2.5 exposure was 6% (95% CI 4, 8%) for all-cause and 11% (95% CI 5, 16%) for cardiovascular mortality. Long-term exposure to PM2.5 was more associated with mortality from cardiovascular disease (particularly ischemic heart disease) than from non-malignant respiratory diseases (pooled estimate 3% (95% CI −6, 13%)). Significant heterogeneity in PM2.5 effect estimates was found across studies, likely related to differences in particle composition, infiltration of particles indoors, population characteristics and methodological differences in exposure assessment and confounder control. All-cause mortality was significantly associated with elemental carbon (pooled estimate per 1 μg/m3 6% (95% CI 5, 7%)) and NO2 (pooled estimate per 10 μg/m3 5% (95% CI 3, 8%)), both markers of combustion sources. There was little evidence for an association between long term coarse particulate matter exposure and mortality, possibly due to the small number of studies and limitations in exposure assessment. Across studies, there was little evidence for a stronger association among women compared to men. In subjects with lower education and obese subjects a larger effect estimate for mortality related to fine PM was found, though the evidence for differences related to education has been weakened in more recent studies.

Keywords: Air pollution, Mortality, Motorized traffic, Cardiovascular, Respiratory, Particles

Review

Background

There is growing evidence of mortality effects related to long-term exposure (i.e., exposures of a year or more) to ambient air pollution [1-3]. Cardiovascular effects of short- and long-term exposure to particulate matter air pollution focusing on PM2.5 have recently been comprehensively reviewed [4,5]. Experimental and epidemiological studies in the recent decade have significantly increased our knowledge of mechanisms that could plausibly explain the associations observed in epidemiological studies between ambient air pollution and mortality [4].

Most studies have reported associations linked to particulate matter, often represented by the mass concentration of particles smaller than 10 μm (PM10) or 2.5 μm (PM2.5). In many urban areas, motorized traffic emissions are an important source of ambient particles and gaseous pollutants such as nitrogen oxides (NO2 and NO). Exposure contrasts related to traffic emissions are usually poorly represented by the concentration of PM10.
or PM$_{2.5}$, because of the high regional background concentration of these particle metrics from other sources [6,7]. However, there are more specific markers for traffic related air pollution, which include elemental carbon and ultrafine particles number [7-10]. Janssen and co-workers recently demonstrated that health impact assessments of traffic-related pollutants based upon PM$_{2.5}$ seriously underestimated the health risks compared to an assessment based upon elemental carbon [7]. There is also growing evidence of health effects related to ultrafine particles [8,9]. Finally, the effects of coarse particles (the particle fraction between 2.5 and 10 μm) have attracted renewed attention [11]. Emission controls for road traffic have now substantially reduced tailpipe emissions, and therefore non-tailpipe emissions including engine crankcase emissions (combusted lubricating oil), road, tire and brake wear are becoming increasingly important. A recent study in the Netherlands found similar increases of concentrations in major roads compared to urban background for metals related to break and tire wear (Cu, Zn) as for soot and ultrafine particles which are due to tailpipe emissions [10]. In a review of the limited literature, coarse particles were associated with short-term effects on mortality and hospital admissions, but no evidence was found for long-term exposure effects [11]. The number of studies on long-term coarse particle exposure reviewed was small however at the time.

The aim of the current review is to evaluate the epidemiological evidence for cardiovascular and respiratory mortality effects of long-term exposure to fine particulate matter, including a meta-analysis. We focused on epidemiological studies of mortality, as experimental studies and mechanisms of effect have been discussed in detail previously [4]. The American Heart Association review [4] is updated with a significant number of new studies published in 2009 – 2012. We further include more pollutants in the review, specifically NO$_2$, elemental carbon and coarse particles. We evaluated the findings on potentially susceptible subgroups across studies of PM$_{2.5}$. In addition, we have included the studies on more specific cardiovascular causes of death, especially fatal myocardial infarction and stroke.

**Methods**

We performed a search in the databases Medline and Scopus with the search terms air pollution, cohort, and mortality until January 2013. We supplemented the search with studies included in the review by Brook and co-worker [4] and by browsing the reference lists of identified papers. In case more than five studies were identified, we performed a meta-analysis. We tested for heterogeneity of cohort-specific effect estimates and obtained combined effects estimates, using random effects methods of DerSimonian and Laird [12]. The I$^2$ statistic was calculated as a measure of the degree of heterogeneity across studies [13]. I$^2$ ranges from 0 to 100% and can be interpreted as the variability of study-specific effect estimates attributable to true between study effects. From some studies multiple papers were available such as the Six Cities study [14-16]. In the meta-analysis we used only the most recent paper, which had longer follow-up. We only included studies in the quantitative meta-analysis that directly provided PM$_{2.5}$ exposure estimates. For NO$_2$ we only included studies which accounted for intra-urban spatial variation using e.g. dispersion models, land use regression models or spatial interpolation. We used STATA version 10 (Stata Corp, College Station, Texas) for meta-analysis. Effect estimates are presented as excess risks expressed per 10 μg/m$^3$ contrast in exposure, except elemental carbon for which risks were expressed per 1 μg/m$^3$.

**PM$_{2.5}$ and all-cause and cardiovascular mortality**

Table 1 and Figures 1 and 2 summarize the studies on long-term air pollution exposure and all-cause and cardiovascular mortality using PM$_{2.5}$ or PM$_{10}$ as exposure metric [14-39]. Most but not all studies report significant associations between PM$_{2.5}$ and all-cause mortality. Since the publication of the authoritative American Heart Association Scientific Statement, sixteen new cohort studies were published between 2009 and January 2013. These studies were often performed in more selected groups e.g. female teachers [27,36] or male truck drivers [32]. The geographic range has also been expanded significantly with several new studies from Japan and China now published. Another tendency is the publication of large studies based upon large population samples (e.g. census), with often less information on confounding variables such as individual smoking habits. Large cohort studies have used neighborhood socio-economic status and co-morbidities strongly associated with smoking as proxies for actual smoking data [26,38]. Effect estimates differed substantially across studies, with most studies showing less than 10% increase in mortality for an increment of 10 μg/m$^3$ PM$_{2.5}$. The random effects summary estimate for the percent excess risk per 10 μg/m$^3$ PM$_{2.5}$ for all-cause mortality was 6.2% (95% CI: 4.1 – 8.4%). A formal test of heterogeneity was statistically significant, with an I$^2$ value of 65% indicating moderate heterogeneity. I$^2$ can be interpreted as the variability in effect estimates due to true between study variability and not chance [13]. The random effects summary effect estimate for cardiovascular mortality was 10.6% (95% CI 5.4, 16.0%) per 10 μg/m$^3$. Thus, the overall effect estimates were larger for cardiovascular than for all-cause mortality. This pattern was found in most of
Table 1 Summary of effect estimates (excess risk per 10 μg/m³) from cohort studies on particulate matter (PM10 or PM2.5) and mortality from all causes and cardiovascular diseases

| Study                                      | Study population                             | Follow-up period | Pollutant | Conc \(^a\) (μg/m³) | Spatial scale\(^b\) | % change in risk (95% CI) | References |
|--------------------------------------------|------------------------------------------------|------------------|-----------|----------------------|----------------------|---------------------------|-------------|
| Harvard six cities                         | 8111 adults in six US cities                 | 1976 - 1989      | PM\(_{2.5}\) | 18 (11–30)           | City                 | 13 (4, 23)                | 18 (6, 32)   | [15]        |
| Harvard six cities                         | 8096 adults in six US cities                 | 1979 -1998       | PM\(_{2.5}\) | 15 (10–22)           | City                 | 16 (7, 26)                | 28 (13,44)  | [14]        |
| Harvard six cities                         | 8096 adults in six US cities                 | 1974 - 2009      | PM\(_{2.5}\) | 16 (11–24)           | City                 | 14 (7, 22)                | 26 (14, 40)  | [16]        |
| American Cancer Society (ACS) study        | 552, 800 adults from 51 US cities            | 1982 - 1989      | PM\(_{2.5}\) | 18 (9–34)            | City                 | 26 (8, 47)                | NA          | [17]        |
| ACS study                                  | 500,000 adults from 51 US cities             | 1982 -1998       | PM\(_{2.5}\) | 18 (4)               | City                 | 6 (2, 11)                 | 9 (3, 16)   | [18]        |
| ACS sub-cohort study                       | 22,905 subjects in Los Angeles area         | 1982 - 2000      | PM\(_{2.5}\) | (<9 – 27)            | Zip code (Int)       | 17 (5, 30)                | 26 (1, 60)  | [19]        |
| German cohort                              | 4752 women in Ruhr area                      | 1985 – 2003      | PM\(_{10}\)  | 44 (35–53)           | Address (near)       | 12 (–9, 37)               | 52 (8, 114) | [20]        |
| German cohort                              | 4752 women in Ruhr and surrounding area     | 1985 – 2008      | PM\(_{10}\)  | 44 (35–53)           | Address (near)       | 22 (6, 41)                | 61 (26, 104)| [21]        |
| Women’s Health Initiative Observational Study | 65,893 postmenopausal women from 36 US metropolitan areas | 1994-1998      | PM\(_{2.5}\) | 14 (3–28)            | Zip code (near)      | NA                        | 76 (25,147) | [22]        |
| Netherlands Cohort Study                   | 120, 852 subjects from Netherlands           | 1987 -1996       | PM\(_{2.5}\) | 28 (23–37)           | Address (LUR)        | 6 (–3, 16)                | 4 (–10, 21) | [23]        |
| Nurses’ Health Study                       | 66,250 women from the US north eastern metropolitan areas | 1992-2002      | PM\(_{10}\)  | 22 (4)               | Address (LUR)        | 11 (1,23)                 | 35 (3, 77)  | [24]        |
| Nurses’ Health Study                       | 66,250 women from the US north eastern metropolitan areas | 1992-2002      | PM\(_{2.5}\) | 14 (6–28)            | Address (LUR)        | 26 (2, 54)                | NA          | [25]        |
| Medicare national cohort                   | 13.2 million elderly Medicare recipients across the USA | 2000 - 2005    | PM\(_{2.5}\) | 13 (4)               | Zip code (Mean)      | 4 (3, 6)                  | [26]        |
| California teachers study                  | 45,000 female teachers                       | 2002 -2007       | PM\(_{2.5}\) | 18 (7–39)            | Address (near)       | 6 (–4, 16)                | 19 (5, 36)  | [27]        |
| Swiss national cohort                      | National census data linked with mortality   | 2000 - 2005      | PM\(_{10}\)  | 19 (>40)             | Address (Disp)       | NA                        | –1 (~3, 0)  | [28]        |
| Health professionals follow-up study       | 17,545 highly educated men in the midwestern and northeastern US | 1989 – 2003    | PM\(_{2.5}\) | 18 (3)               | Address (LUR)        | –14 (~28,2)               | 3 (~17, 26) | [29]        |
| Vancouver cohort                           | 452,735 Vancouver residents 45–85 yr        | 1999 – 2002      | PM\(_{2.5}\) | 4 (0 – 10)           | Address (LUR)        | NA                       | 7 (~14, 32) | [30]        |
| China nat. hypertension survey             | 70,497 men and women                         | 1991 - 2000      | TSP        | 289 (113–499)        | City                 | 0.3 (0, 1)                | 1 (0, 2)    | [31]        |
| US trucking industry cohort                | 53,814 men in the US trucking industry       | 1985 -2000       | PM\(_{2.5}\) | 14 (4)               | Address (near)       | 10 (3, 18)                | 5 (~7, 19) | [32]        |
| Chinese retrospective cohort study         | 9,941 adults from five districts of Shenyang city | 1998 -2009      | PM\(_{10}\)  | 154 (78–274)         | District (mean)       | 53 (50, 56)               | 55 (51, 60) | [33]        |
| Canadian national cohort                   | 2.1 million nonimmigrant Canadians . > 25 yr | 1991 - 2001      | PM\(_{2.5}\) | 9 (2 – 19)           | Enumeration area, N = 45710 (satellite) | 10 (5, 15)                | 15 (7, 24) | [34]        |
| New Zealand Census mortality study         | 1.06 million adults in urban areas from 1996 census | 1996 -1999      | PM\(_{10}\)  | 8 (0 – 19)           | Census tract (Disp)   | 7 (3, 10)                 | 6 (1, 11)  | [35]        |
Table 1 Summary of effect estimates (excess risk per 10 μg/m³) from cohort studies on particulate matter (PM₁₀ or PM₂.₅) and mortality from all causes and cardiovascular diseases (Continued)

| Study                          | N/A                  | Year       | PM₁₀/PM₂.₅ | Address (Method) | Excess Risk (Median, 90th percentile) | Ref. |
|--------------------------------|----------------------|------------|-------------|------------------|--------------------------------------|------|
| California teachers study      | 101,784 female teachers | 1997-2005  | PM₂.₅      | Address (Inter)  | 16 (3–28) 7 (–5, 19)                 | [36] |
| Nippon data cohort             | 7,250 adults > 30 yr throughout Japan | 1980-2004 | PM₁₀       | District (near)  | −2 (–8, 4) −10 (–19, 0)              | [37] |
| Rome longitudinal study        | 1,265,058 adults from Rome | 2001-2010  | PM₂.₅      | Address (DISP, 1 km grid) | 23 (7 – 32) 4 (3, 5) 6 (4, 8)         | [38] |

a Mean with minimum – maximum in parentheses (μg/m³). One number in parentheses is standard deviation.

b Spatial scale of exposure assignment, in parentheses exposure assignment method. City = average of monitors within the city; Near = nearest monitor concentration; LUR = land use regression; Disp = dispersion modeling; Inter = interpolation.

c Cardio-pulmonary mortality reported if cardiovascular mortality not available.
d Combining the estimates from the three regions of the USA.
e Median and 90th percentile reported.

f Very high pollution levels that changed significantly during follow-up changing the ranking of the five districts.

Studies adjusted for individual smoking except references [26,28,30,34,38,56].
the individual studies, with the exceptions being the Dutch cohort study [23], the US trucking industry cohort study [32] and a national cohort study from New Zealand [35]. Significant and large heterogeneity of effects was found across studies, with an I² statistic of 61%. After excluding the Miller study [22], moderate heterogeneity remained (I² = 40%). Overall, the new studies have supported an association between PM 2.5 and mortality first identified in the US Six City and ACS studies. It is of interest to note that the weight of the ACS study in the combined effect estimate is 12% for all-cause mortality, documenting that the combined estimate does not rely on one or two studies. Furthermore, effect estimates from the three large population cohorts without individual smoking data [26,34,38] were not higher than those from the individual cohort studies. An important question is what the explanation is for the observed heterogeneity of effect estimates. Differences in study population, exposure assessment, pollution mixture, study period, outcome assessment, and confounder control could have contributed to these differences. 

Effect modification

Differences in the fraction of susceptible subjects may have contributed to the observed differences. Brook [4] suggested that women might be more susceptible to ambient air pollution. The studies with higher PM effect estimates, particularly the WHI-study have indeed been performed in women only. However, it is problematic to draw conclusions about susceptible subgroups based upon between-study comparisons as multiple factors differ between studies. A comparison of PM effect estimates between men and women within studies does not provide clear evidence that women have a stronger response (Table 2). The findings from the AHSMOG are difficult to interpret, with higher effects in men in the larger earlier study [40] and larger effects in women in the smaller cohort with longer follow-up [41]. The larger effect estimate for BC for men in a Canadian study [30] has to be interpreted with care, because of the lack of data on a variety of important covariates, including individual smoking data, though the authors argue that smoking likely has not confounded the associations with mortality. In the French PAARC study, effect estimates for the evaluated pollutants
(TSP, BS and NO₂) were similar among men and women [42]. There is also only weak evidence that effect estimates are larger among never-smokers, though in all evaluated studies a (borderline) significant association was found in never-smokers (Table 2). Associations in current smokers were more variable across the studies, consistent with the larger ‘noise’ generated by smoking. In all four studies, PM₂.₅ effect estimates were higher for those with the lowest education and there was little indication of an association in those with higher education. The absence of an association in the (highly educated) Health professionals study [29] is consistent with this observation. In contrast, in the French PAARC study, effect estimates for Black Smoke were very similar across educational strata, with significant effect also found in those with a university degree [42]. Furthermore the PM₂.₅ effect estimates (excess risks) in an extended analysis of the ACS differed less than originally reported: 8.2%, 7.2 and 5.5% per 10 μg/m³ for subjects with low, medium and high education respectively [43]. If confirmed in further studies, it is likely that multiple life style related factors may play a role in the stronger effects observed in less-educated subjects. These may include dietary factors such as lower fruit and antioxidant intake [23], higher risk of obesity or other pre-existing diseases, higher actual exposures than assumed in the studies, lack of air conditioning and possibly interaction with other risk factors such as poorer housing conditions e.g. moisture.

In two studies, PM₂.₅ effect estimates were substantially higher among subjects with high body mass index [22,24]. It is likely that subject characteristics might explain part of the variability of air pollution effect estimates across studies where subgroup analyses are limited by power to detect differences. Hence, further research is required to study the effects of air pollution on women, smokers, obese participants, and diabetes mellitus with better measurement of the exposures. Gene-environment interactions have been shown for the (short-term) air pollution effects on inflammation markers [44,45] Inflammation likely plays an important role in the mechanism of cardiovascular events [3,4]. Gene-environment interactions have not yet been studied in the framework of mortality cohort studies.

**Figure 2** Meta-analysis of the association between PM₂.₅ and cardiovascular mortality (Relative Risk per 10 μg/m³). Overall uses random effects.
**Exposure issues**

One of the important sources of variability of effect estimates between studies is likely related to exposure definition and misclassification. While the most important environmental predictor to consider is actual individual-level exposure to ambient particles, which presumably drives the health effects, most studies have used outdoor concentrations at sites distant to the participant’s precise location. The use of outdoor exposures leads to exposure misclassification. In the cohort studies, exposure has been characterized by the outdoor concentration at the city level based upon central site monitoring or the nearest monitor, or modeling at the individual address. Table 1 shows that the spatial scale of assessment and exposure assessment method varied significantly across studies, probably contributing to differences in effect estimates. Differences in pollution range across studies (Table 1) may have contributed as well. These exposure estimates do not take into account time activity patterns such as time spent in the home or in traffic and factors affecting infiltration of particles indoors. There is a large literature documenting the importance of air exchange rate on infiltration of particles indoors. Importantly, these factors may differ between homes within a study area and between study areas in different climates. In a study of short-term effects, PM10 effects on hospital admissions were larger in US cities with lower% of air conditioning, related to higher particle infiltration rates [46]. The impact of air conditioning use has not been investigated yet in the framework of cohort studies. In the Multiethnic study of Atherosclerosis Air study, indoor-outdoor measurements have been performed to adjust the exposure estimates [47,48] and each participant provides time-activity information to weight exposures between time spent indoors and outdoors. Evidence for the importance of time activity patterns was obtained in the US truckers study, showing higher ambient PM2.5 effect estimates in the population excluding long-haul drivers who spend more time away from home [32]. Other factors could however also explain the higher effect estimated after excluding long-haul drivers. In the WHI study, effect estimates tended to be higher for subjects spending more than 30 minutes outdoors [22]. In a validation study in the Netherlands, the contrast of personal soot exposure for adults living on a major road compared to those living at a background location, was larger for those spending more time at home [49]. Because of the reliance on ambient exposure estimates, it is not surprising that some heterogeneity in effect estimates across studies is found.

Differences in particle composition or contributing sources very likely explain some of the heterogeneity in effect estimates, as was observed for short-term mortality and hospital admission studies of PM2.5 and PM10 [50-53]. For a comprehensive review we refer to the recent evaluation made by the World Health Organization (http://www.euro.who.int/en/what-we-do/health-topics/environment-and-health/air-quality/publications/2013/review-of-evidence-on-health-aspects-of-air-pollution-revhaap). Particle composition effects have not been systematically investigated in cohort studies with the exception of the California teacher’s study [27]. In a recent review it was shown that on a per microgram per m³
basis, mortality effect estimates were about 10 times larger for EC than for PM$_{2.5}$ [7]. Hence, in locations with higher levels of primary combustion particles we could expect higher PM$_{2.5}$ effects. In the next section, evidence on EC is further discussed.

A further important issue is for which period exposure is characterized. Air pollution data may not be available for the entire follow-up period. As an example in the ACS study, PM$_{2.5}$ data were available at the start and end of follow-up [18]. When significant (often downward) trends in pollution occur with changing (often decreasing) spatial contrasts in the study, bias may occur in the estimated association between pollution and mortality. The follow-up study from the Harvard Six City study [14] and two studies in potentially at-risk populations [54,55] suggested that the relevant exposure for cardiovascular effects may be the exposure in the past few years. These authors conclude that it does not take decades to bridge the gap between the short- and long-term exposure effect estimates, consistent with the effect of intervention studies showing reductions in mortality in the year after the intervention [54,55]. These studies [54,55] have made use of long-term temporal contrast within cities adjusting for secular trends. PM effect estimates were similar to the previously discussed studies exploiting spatial contrasts.

A further temporal issue in studies that use land use regression models for exposure assessment is that these models often are based upon current measurement campaigns and linked to health outcomes that occurred in the past. Three studies in the Netherlands, Rome (Italy) and Vancouver (Canada) have shown that for periods of about 10 years current LUR models predicted historic spatial contrasts well [56-58]. Even when concentrations have decreased over time, spatial contrasts often remain stable. Spatial contrasts may not be stable in areas with rapid economic development as indicated in one of the Chinese cohort studies in which the ranking of study areas changed during follow-up [33,59]. Even when the ranking of subjects is not changed, the quantitative spatial contrast in a study area may have changed, e.g. because the difference between major roads and background locations has decreased in time. Changed spatial contrasts will affect the estimated slope of the mortality pollution association [18,56]. Moving of subjects may further complicate the assessment.

An important question to address for the traffic pollution studies is potential confounding by road traffic noise, which has been shown to be related to cardiovascular disease including MI as well. A few studies have attempted to disentangle traffic-related air pollution and noise [60-62]. These studies found moderate correlations between air pollution and noise. The three studies differed somewhat in their findings of independent air pollution and noise effects. More work is needed in this area.

**Coarse particles and elemental carbon**
Table 3 presents studies that have used elemental carbon or coarse PM as the exposure metric. Table 3 illustrates that there is no evidence that long-term exposure to coarse PM is related to mortality. In three of the four cohort studies that reported no significant association with coarse PM, significant associations with PM$_{2.5}$ were found [18,25,63]. However, exposure assessment for coarse particles is more challenging than for PM$_{2.5}$ because of the influence of local sources, hence central site monitors are likely to have greater errors in representing residential concentrations. It is therefore possible that with more spatially resolved exposure assessment methods such as land use regression models or dispersion models, potential long-term exposure effects will be detected. The California Teacher’s study did not evaluate coarse PM and did not find significant associations between all-cause mortality and elemental concentrations of Si, Fe and Zn, elements abundant in coarse particles, but did report an association between Si and ischemic heart disease [27].

Consistently, the summary estimate for PM$_{10}$ was smaller than for PM$_{2.5}$ with a summary effect estimate per 10 μg/m$^3$ of 3.5% (95% CI 0.4%, 6.6%) with significant heterogeneity ($I^2 = 69\%$) of the studies included in Table 1, excluding the because of changing spatial patterns difficult to interpret Chinese retrospective study [33]. The PM$_{10}$ analysis was added as several studies only report PM$_{10}$.

Effect estimates for EC were very consistent across studies [23,27,30,42,64-67]. The random effects summary estimate for all-cause mortality per 1 μg/m$^3$ EC was 6.1% (95% CI 4.9%, 7.3%), with highly non-significant heterogeneity of effect estimates ($I^2 = 0\%$). Most of the included studies assessed EC exposure at the city-scale [27,64] which represents variation in city background but does not account for small-scale variation related to proximity to major roads. Many studies have documented significant intra-urban contrasts for EC, related to especially major roads [7]. Most likely EC and NO$_2$ should be considered representatives of the complex mixture of traffic-related air pollution, rather than the only components causally associated with mortality.

There is fairly consistent evidence of associations of mortality with nitrogen dioxide (Table 4). The random effects summary estimate for all-cause mortality per 10 μg/m$^3$ for NO$_2$ was 5.5% (95% CI 3.1%, 8.0%), with significant and large heterogeneity of effect estimates ($I^2 = 73\%$). In this analysis, the Chinese study [33] was not included as exposure was assessed at the district level. Inclusion of the essentially null findings of the ACS study—excess risk of 0.3% (95% CI −0.8, 1.3%)- resulted in an only
Table 3 Summary of effect estimates (excess risk per 10 $\mu$g/m$^3$) from cohort studies on coarse particulate matter and elemental carbon (per 1 $\mu$g/m$^3$) and mortality from all causes and cardiovascular diseases

| Study name                  | Study design                                                                 | Follow-up period     | Pollutant | Conc$^a$ ($\mu$g/m$^3$) | Spatial scale$^b$ | % change in risk (95% CI) in mortality | References |
|-----------------------------|-----------------------------------------------------------------------------|----------------------|-----------|--------------------------|------------------|----------------------------------------|------------|
| Coarse PM                   |                                                                             |                      |           |                          |                  |                                        |            |
| ACS study                   | 500,000 adults 51 US cities                                                 | 1982 - 1998          | PM$_{2.5-15}$ | 19 (6)                  | City             | 1 (−2.3) 2 (−2, 5)$^*$                 | [18]       |
| AHSMOG study                | 3769 California seventh-day Adventists                                       | 1977 – 1992          | PM$_{2.5-15}$ | 27 (4 – 44)             | Address (Inter)  | 5 (−8, 20) NA                          | [63]       |
| Nurses’ Health Study        | 66,250 women from US north eastern metropolitan areas                       | 1992- 2002           | PM$_{0.5-10}$ | 8 (0 – 27)              | Address (LUR)    | 3 (−11, 18) NA                         | [25]       |
| Health professionals follow-up study | 17,545 highly educated men in the midwestern and northeastern US              | 1989 – 2003         | PM$_{0.5-10}$ | 10 (3)                  | Address (LUR)    | −10 (−22, 4) 8 (−10, 29)              | [29]       |
| EC                          |                                                                             |                      |           |                          |                  |                                        |            |
| Netherlands Cohort Study    | 120, 852 subjects from Netherlands                                           | 1987 - 1996          | BS$^a$    | 17 (9–36)               | Address (LUR)    | 5 (0, 11) 4 (−5, 13)                   | [23]       |
| ACS study (extended)        | 500,000 adults 51 US cities                                                 | 1982 – 1998          | EC        | IQR = 0.31              | City             | 6 (1, 11) 11 (3, 19)                   | [64]       |
| Worcester MI survivors      | 3,895 MI patients                                                           | 1995 - 2005          | EC        | 0.4 (0.1 – 0.9)         | Address (LUR)    | 2 (−7, 11)$^d$ NA                      | [65]       |
| Vancouver cohort            | 452,735 Vancouver residents 45–85 yr                                       | 1999 – 2002          | BC        | 1.5 (0–5)               | Address (LUR)    | NA                                     | [30]       |
| PAARC                       | 14,284 adults in 24 French areas                                            | 1974 – 1998          | BS        | 44 (18–77)              | Address (near)   | 7 (3, 10) 5 (−2, 12)                   | [42]       |
| Veteran’s study             | 70,000 male US veterans                                                    | 1997 – 2001          | EC        | 0.6 (0.1 – 2.0)         | County (mean)    | 18 (5, 33) NA                          | [66]       |
| California teachers study   | 45,000 female teachers                                                      | 2002 - 2007          | EC        | 1.1 (0.2 – 2.4)         | Address (near)   | 3 (−11,19) 11 (−9, 36)                | [27]       |
| Two Scotch cohorts          | 15, 402 and 7,028 adults from West-central and central Scotland              | 1972 – 1998 1970 - 1998 | BS    | 19                      | LUR + temporal   | 5 (1,9) 7 (0, 13)                      | [67]       |

$^a$ Mean with minimum – maximum in parentheses ($\mu$g/m$^3$). One number in parentheses is standard deviation.

$^b$ Spatial scale of exposure assignment, in parentheses exposure assignment method. City = average of monitors within the city; Near = nearest monitor concentration; LUR = land use regression; Disp = dispersion modeling; Inter = interpolation.

$^c$ Cardio-pulmonary mortality reported if cardiovascular mortality not available.

$^d$ HRs for first two years after MI and after the first two years of survival.

$^*$ BC (Black Carbon), BS (Black Smoke) and EC (Elemental Carbon) are different markers used to assess soot. Increases consistent with a 1 $\mu$g/m$^3$ increase in EC were used [7].

Studies adjusted for individual smoking except references [26,28,30,34,38,56].
Table 4 Summary of cohort studies on NO₂ and mortality from all causes and cardiovascular diseases (excess risk per 10 μg/m³)

| Study name                      | Study population                          | Follow-up period | Pollutant | Conc³ (µg/m³) | Spatial scaleᵇ | % change in risk (95% CI) in mortality per 10 µg/m³ | References |
|---------------------------------|-------------------------------------------|------------------|-----------|---------------|----------------|-----------------------------------------------------|------------|
| Oslo cohort                     | 16,209 men in Oslo, Norway                | 1972 – 1998      | NOₓ       | 11 (1 – 168)  | Address (DISP) | 8 (6,11) NA                                          | [68]       |
| Netherlands Cohort Study        | 120,852 subjects from Netherlands         | 1987 -1996       | NO₂       | 37 (15–67)    | Address (LUR) | 8 (0, 16) 7 (–6, 21)                                 | [23]       |
| German cohort                   | 4752 women in Ruhr and surrounding area  | 1985 – 2003      | NO₂       | 39 (20 – 60)  | Address (near) | 11 (1,21) 36 (14, 63)                                | [20]       |
| German cohort                   | 4752 women in Ruhr and surrounding area  | 1985 – 2008      | NO₂       | 39 (20 – 60)  | Address (near) | 11 (4,18) 32 (18, 47)                                | [21]       |
| PAARC                           | 14,284 adults in 24 French areas          | 1974 – 1998      | NO₂       | 39 (12 – 32)  | Address (near) | 14 (3, 25) 27 (4, 56)                                 | [42]       |
| China nat. hypertension survey  | 70,497 men and women                      | 1991 - 2000      | NOₓ       | 50 (20 – 122) | City           | 2 (0, 3) 2 (1, 4)                                    | [31]       |
| Vancouver cohort                | 452,735 Vancouver residents aged 45–85 yr| 1999 – 2002      | NO₂       | 32 (15 – 58)  | Address (LUR) | NA 5 (1, 9)                                          | [30]       |
| DCH                             | 52,061 adults in Copenhagen and Arhus     | 1993 - 2009      | NO₂       | 17 (11 – 60)  | Address (DISP) | 8 (2, 13) 15 (3,27)                                  | [69]       |
| US trucking industry cohort     | 53,814 men in the US trucking industry    | 1985 -2000       | NO₂       | 28 (14)       | Address (LUR) | 5 (3, 7) 4 (0, 8)                                    | [32]       |
| Chinese retrospective cohort study| 9,941 adults from five districts of Shenyang city | 1998 -2009 | NO₂ | 46 (18–78) | District (mean) | 145 (134, 158) 146 (131, 163) | [33] |
| Rome longitudinal study         | 684,000 adults from Rome                  | 2001 - 2006      | NO₂       | 45 (11)       | Address (LUR) | 4 (3, 5) NA                                          | [56]       |
| California Teachers study       | 101,784 female teachers                   | 1997 -2005       | NO₂       | 67 (10 – 134) | Address (Inter) | –3 (–9, 4) –2 (–12, 9)                                | [36]       |
| Shizuoka elderly cohort         | 13,444 adults > 65 yr                     | 1999 - 2006      | NO₂       | 25 (–19, 75)  | Address (LUR) | 2 (–4, 8) 15 (3, 28)                                 | [70]       |
| Ontario tax cohort              | 205, 440 adults in Toronto, Hamilton/Windsor| 1982 – 2004 | NO₂ | 43 (8), 31 (6), 24 (5³) | Address (LUR) | NA 8 (5, 11) 8 (5, 11) | [71] |
| Rome longitudinal study         | 1,265,058 adults from Rome                | 2001 - 2010      | NO₂       | 44 (13–75)    | Address (LUR) | 3 (2, 3) 3 (2, 4)                                    | [38]       |

¹ Mean with minimum – maximum in parentheses (µg/m³). One number in parentheses is standard deviation.
² Spatial scale of exposure assignment, in parentheses exposure assignment method. City = average of monitors within the city; Near = nearest monitor concentration; LUR = land use regression; Disp = dispersion modeling; Inter = interpolation.
³ Mean (IQR) per city.

Studies adjusted for individual smoking except references [26,28,30,34,38,56].
Table 5 Summary of the studies on particulate matter and NO$_2$ and mortality from specific cardiovascular diseases (excess risk per 10 μg/m$^3$)

| Study name                        | Pollutant | Conc$^a$ (μg/m$^3$) | Spatial scale$^b$ | % change in risk (95% CI) in mortality associated with a 10 μg/m$^3$ increase | References |
|-----------------------------------|-----------|---------------------|-------------------|---------------------------------------------------------------------------------|------------|
|                                   |           |                     |                   | IHD mortality                                                                     | M.I. mortality | Cerebrovascular mortality |
| ACS study                         | PM$_{2.5}$ | 17 (5)              | City              | 18 (14, 23)                                                                      | NA          | 2 (−5, 10)               | [39]    |
| Oslo cohort                       | NO$_x$    | 11 (1 – 168)        | Address (DISP)    | 8 (3, 12)                                                                        | NA          | 4 (−6, 15)               | [68]    |
| Women’s Health Initiative Study   | PM$_{2.5}$ | 14 (3–28)           | Zip code 5 (near) | 76 (25,147)                                                                      | NA          | NA                       | [22]    |
| Netherlands Cohort Study          | BS        | 17 (9–36)           | Address (LUR)     | 1 (−17, 22)                                                                      | NA          | 39 (−1, 94)              | [23]    |
| Nurses’ Health Study              | PM$_{10}$ | 22 (4)              | Address (LUR)     | 35 (3, 77)                                                                       | NA          | NA                       | [24]    |
| Nurses’ Health Study              | PM$_{2.5}$ | 14 (6–28)           | Address (LUR)     | NA                                                                              | 102 (7, 278) | NA                       | [25]    |
| California teachers study         | PM$_{2.5}$ | 18 (7–39)           | Address (near)    | 55 (24, 93)                                                                      | NA          | NA                       | [27]    |
| Swiss national cohort             | PM$_{10}$ | 19 (>40)$^c$        | Address (Disp)    | −1 (−3, 0)                                                                       | NA          | −1 (−2, 0)               | [28]    |
| Health professionals follow-up study | PM$_{2.5}$ | 18 (3)              | Address (LUR)     | −2 (−30, 35)                                                                     | NA          | NA                       | [29]    |
| Canadian national cohort          | PM$_{2.5}$ | 9 (2 – 19)          | Enumeration area, N = 45710 (satellite) | 30 (18,43)                                                                        | NA          | 4 (−7, 16)               | [34]    |
| Californian Teachers study        | PM$_{2.5}$ | 16 (3–28)           | Address (Inter)   | 20 (2, 41)                                                                        | NA          | 16 (−8, 46)              | [36]    |
| Shizuoka elderly cohort           | NO$_2$    | 25 (−19, 75)        | Address (LUR)     | 27 (2, 58)                                                                       | NA          | 9 (−6, 27)               | [70]    |
| Nippon data cohort                | PM$_{10}$ | <27 - > 43          | District (near)    | −8 (−27, 17)                                                                     | NA          | −14 (−26, 1)             | [37]    |
| DCH                               | NO$_2$    | 17 (11 – 60)        | Address (Disp)    | 7 (−9, 26)                                                                       | NA          | 6 (−14, 32)              | [69]    |
| Ontario Tax cohort                | NO$_2$    | 43 (8), 31 (6), 24 (5)$^c$ | Address (LUR)      | 9 (4, 14)                                                                        | NA          | −4 (−10, 5)              | [71]    |
| Rome longitudinal study           | PM$_{2.5}$ | 23 (7 – 32)         | Address (DISP, 1 km grid) | 10 (6, 13)                                                                        | NA          | 8 (4, 13)                | [38]    |
| M.I. registry studies             |           |                     |                   |                                                                                 |             |                          |         |
| Stockholm                         | NO$_2$    | 14 (3 – 32)         | Address (DISP)    | NA                                                                              | 15 (−1, 33) | NA                       | [72]    |
| Rome residents                    | NO$_2$    | (<30 - > 60)        | Census block (LUR) | NA                                                                              | 7 (2, 12)   | NA                       | [73]    |
| Stockholm residents               | NO$_2$    | 12 (2 – 33)         | Address (DISP)    | NA                                                                              | 8 (5, 11)   | NA                       | [74]    |

IHD = ischemic heart disease; M.I. = myocardial infarction. Fatal M.I. reported for registry studies. NA = not available.

$^a$ Mean with minimum – maximum in parentheses (μg/m$^3$). One number in parentheses is standard deviation.

$^b$ Spatial scale of exposure assignment, in parentheses exposure assignment method. City = average of monitors within the city; Near = nearest monitor concentration; LUR = land use regression; Disp = dispersion modeling; Inter = interpolation.

$^c$ Median and 90th percentile reported.

Studies adjusted for individual smoking except references [26,28,30,34,38,56].
slightly smaller combined estimate of 4.7% (95% CI 2.4, 7.1%). In the ACS study, intra-urban variation was also not accounted for. As traffic-related air pollution varies on a small spatial scale, it is even more critical to assess exposure on a fine spatial scale such as the residential address than for PM2.5.

Specific cardiovascular causes of death
Table 5 shows associations between ambient air pollution and mortality from ischemic heart disease or myocardial infarction (MI), including studies based upon death certificates, more detailed studies using registry data, or ideally cohort studies with epidemiological review of medical records, allowing more precise identification of disease incidence. Several case–control studies based upon MI registries or epidemiological studies with clinical review have found associations between NO2 and fatal MI but not non-fatal MI [72–74]. Thus far, the finding of associations for fatal MI only was interpreted as an evidence that air pollution particularly affects the frail, or acts to aggravate a disease progression caused by other factors. On the other hand, it is also possible that the outcomes of ischemic heart diseases are misclassified and combined as composite outcomes, where fatal outcomes are captured more precisely [75]. Although there is increasing evidence that air pollution is associated with markers of early atherosclerosis, it is possible that air pollution will affect the underlying biological processes that predispose to atherothrombosis (which leads to MI and stroke) compared to atherosclerosis [76,77]. Another explanation is that the type of outcomes affected by pollution are those that have higher case-fatality rates (e.g., arrhythmic sudden death has higher case-fatality rate than overall MI).

Fewer studies have evaluated cerebrovascular mortality. In the Dutch cohort study and in the Women’s Health Initiative Study, a strong association was found [22,23]. In contrast, in the ACS study, the Norwegian cohort, and the Swiss national cohort study no association was found [28,39,68]. It is possible that poorer recording of cerebrovascular mortality on death certificates has contributed to these inconsistencies. There is also some evidence from ecological studies that air pollution may contribute to stroke mortality [78,79].

Two studies have reported significant associations between particulate matter air pollution and dysrhythmia, heart failure and cardiac arrest combined [39,60]. These results are based upon smaller numbers of events, and require large cohort studies for further verification. The results are consistent with several studies documenting significant associations between short-term PM or NO2 exposure and mortality due to heart failure and dysrhythmia and defibrillator discharges [4,80].

Table 6 Summary of the studies on air pollution and mortality from all respiratory disease (excess risk per 10 μg/m³)

| Study Name               | Pollutant | Conc<sup>a</sup> (μg/m³) | Spatial scale<sup>b</sup> | % change in risk (95% CI) in mortality per 10 μg/m³ | References |
|--------------------------|-----------|--------------------------|---------------------------|--------------------------------------------------|------------|
| AHSMOG                   | PM<sub>10</sub> | 51 (17)                  | Address (Inter)           | 6 (−1, 15)                                       | [40]       |
| ACS study                | PM<sub>2.5</sub> | 17 (5)                   | City                      | −8 (−14, −2)                                     | [39]       |
| Oslo cohort              | NO<sub>2</sub>  | 11 (1 – 168)             | Address (DISP)            | 16 (6, 26)                                       | [68]       |
| Harvard six cities       | PM<sub>2.5</sub> | 15 (10–22)               | City                      | 8 (−21, 49)                                      | [14]       |
| Netherlands Cohort Study | PM<sub>2.5</sub> | 28 (23–37)               | Address (LUR)             | 7 (−25, 52)                                      | [23]       |
| Netherlands Cohort Study | NO<sub>2</sub>  | 37 (15–67)               | Address (LUR)             | 12 (0, 26)                                       | [23]       |
| California Teachers study| PM<sub>2.5</sub> | 18 (7–39)                | Address (near)            | 3 (−20, 34)                                      | [27]       |
| China national. hypertension survey | NO<sub>2</sub> | 50 (20 – 122)         | City                      | 3 (0, 6)                                         | [31]       |
| China national. hypertension survey | TSP | 289 (113 – 499)   | City                      | 0.3 (−1, 1)                                      | [31]       |
| US truckers study        | PM<sub>2.5</sub> | 14 (4)                   | Address (near)            | 20 (−9, 60)                                      | [32]       |
| US truckers study        | NO<sub>2</sub>  | 28 (14)                  | Address (LUR)             | 15 (1,31)                                        | [32]       |
| California Teachers study| PM<sub>2.5</sub> | 16 (3–28)                | Address (Inter)           | 21 (−3, 52)                                      | [36]       |
| New Zealand Census study | PM<sub>10</sub> | 8 (0 – 19)               | Census tract (Disp)       | 14 (5, 23)                                       | [35]       |
| Shenyang cohort study    | PM<sub>10</sub> | 154 (78 – 274)          | District (mean)           | 67 (60, 74)                                      | [59]       |
| Shenyang cohort study    | NO<sub>2</sub>  | 46 (18–78)               | District (mean)           | 197 (169, 227)                                   | [59]       |
| Shizuoka elderly cohort  | NO<sub>2</sub>  | 25 (−19, 75)             | Address (LUR)             | 19 (2, 38)                                       | [70]       |
| Two Scotch cohorts       | BS         | 19                       | LUR + temporal            | 11 (−3, 28)                                      | [67]       |
| Rome longitudinal study  | PM<sub>2.5</sub> | 23 (7 – 32)             | Address (DISP, 1 km grid) | 3 (−3, 8)                                        | [38]       |

<sup>a</sup> Mean with minimum – maximum in parentheses (μg/m³). One number in parentheses is standard deviation.

<sup>b</sup> Spatial scale of exposure assignment, in parentheses exposure assignment method. City = average of monitors within the city; Near = nearest monitor concentration; LUR = land use regression; Disp = dispersion modeling; Inter = interpolation.

Studies adjusted for individual smoking except references [26,28,30,34,38,56].
Air pollution and respiratory mortality

Table 6 shows the effect estimates for respiratory mortality. In the two first US cohort studies, no association between PM$_{2.5}$ and respiratory mortality was found [15,17]. In contrast to the findings of these US studies, strong associations were found in the Dutch cohort study [23], a Norwegian study [68] and a Chinese study [59]. The random effect pooled per 10 $\mu$g/m$^3$ for PM$_{2.5}$ was 2.9% (95% CI $-5.9$, 12.6%), highly non-significant. The heterogeneity across studies was statistically significant with an $I^2$ statistic of 59%, indicating moderate heterogeneity. Associations for PM were weaker in the Dutch and Chinese cohort study than with NO$_2$ or NO$_x$. Respiratory mortality may be more related to primary traffic-related pollutants than with long-range transported particles, though further work is needed to test this hypothesis. The smaller number of deaths due to respiratory disease compared to cardiovascular diseases, contributed to larger confidence intervals within individual studies and larger variability of the main effect estimates across studies. In time series studies including several large multi-city studies in the USA and Europe, significant associations between daily variations in PM and respiratory mortality were found [1-4]. Expressed per 10 $\mu$g/m$^3$ PM excess risks of about 1% are typically reported for short-term exposures, larger than for all-cause mortality [1-4]. In contrast to cardiovascular disease, current evidence therefore does not suggest an additional risk from long-term exposure, possibly related to mortality displacement [2,3]. More studies are needed to evaluate long-term exposures on respiratory mortality more thoroughly.

Conclusions

There is a significant number of new studies on long-term air pollution exposure, covering a wider geographic area, including Asia. These recent studies support associations found in previous cohort studies on PM$_{2.5}$. The pooled effect estimate expressed as excess risk per 10 $\mu$g/m$^3$ increase in PM$_{2.5}$ exposure was 6% (95% CI 4, 8%) for all-cause and 11% (95% 5, 16%) for cardiovascular mortality. Long-term exposure to PM$_{2.5}$ was more associated with mortality from cardiovascular disease (particularly ischemic heart disease) than from non-malignant respiratory diseases (pooled estimate 3% (95% CI $-2.6$, 13%)). Significant heterogeneity in PM$_{2.5}$ effect estimates was found across studies, likely related to differences in particle composition, infiltration of particles indoors, population characteristics and methodological differences in exposure assessment and confounder control. All-cause mortality was significantly associated with elemental carbon (pooled estimate per 1 $\mu$g/m$^3$ 6% (95% CI 5, 7%)) and NO$_2$ (pooled estimate per 10 $\mu$g/m$^3$ 5% (95% CI 3, 8%)), both markers of combustion sources. There was little evidence for an association between long-term coarse particulate matter exposure and mortality, possibly due to the small number of studies and limitations in exposure assessment. Across studies, there was little evidence for stronger association among women compared to men. Subjects with lower education and obese subjects experienced larger mortality effect related to fine PM, though the evidence for differences related to education has been weakened in more recent studies.

Our review suggests several specific research questions. Research into the reasons for the heterogeneity of effect estimates would be extremely useful for health impact assessment. Better exposure assessment including spatially resolved outdoor exposures and more chemically speciated PM might in part be able to resolve the observed heterogeneity. Chemical speciation would allow assessing particles from different sources e.g. particles from combustion sources and non-tailpipe emissions separately, a question clearly relevant for air pollution control policy. Specific attention to motorized traffic emissions is important because (road) traffic is an important source of ambient air pollution. More work on coarse particles and at the other side of the particle size spectrum, ultrafine particles is needed. Ongoing new research in the USA in the Multi-Ethnic study of Atherosclerosis and Air pollution (MESA-AIR) and the European Study of Cohorts for Air Pollution Effects (ESCAPE) that use large cohorts and state-of-the art spatially-resolved exposure methods will likely contribute significant new answers in the near future to these questions.

Abbreviations

ACS: American Cancer Society study; BS: Black Smoke; BC: Black Carbon; CI: Confidence interval; EC: Elemental Carbon; NO$_2$: Nitrogen dioxide; NO$_x$: Nitrogen oxides; PM: Particulate matter; PM$_{10}$: Particles smaller than 2.5 $\mu$m; PM$_{2.5}$: Particles smaller than 10 $\mu$m; TSP: Total suspended particles.

Competing interests

None of the authors has a competing interest.

Authors’ contributions

GH, RMK, RB, AP, BO, BB and JK have contributed to the definition of the scope of the review, identification of studies and interpretation of results. GH drafted the text. GH, RMK, RB, AP, BO, BB and JK provided critical comments and approved the final manuscript.

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