Appendix Table 1. Exposure metrics, exposure assessment methods, and spatial resolution eligible for inclusion.

Exposure metric	Exposure assessment method	Spatial resolution exposure method	Spatial resolution exposure assignment to participants	Spatial resolution exposure assignment for study identification	Traffic contribution to exposure and other considerations
All pollutants ^a	Dispersion / chemical transport models of traffic emissions or traffic-specific source tracking / apportionment	≤5 km	≤5 km	Residential address as exact address, neighborhood, census tract or block, or postal code (but not city or county)	Assumed by method
All pollutants ^a	Dispersion / chemical transport models of all sources	≤5 km	≤5 km	Residential address as exact address, neighborhood, census tract or block, or postal code (but not city or county)	Judgement needed (e.g., required area adjustment in epidemiological analysis if spatial extent of the study area was >10,000 km², determination of whether exposures met longterm criteria)
All pollutants ^a	Land use regression models that contain at least one traffic predictor (e.g., traffic intensity or road density) or broader surrogates of traffic (e.g., address density, household density, population density, impervious surface)	≤5 km	≤5 km	Residential address as exact address, neighborhood, census tract or block, or postal code (but not city or county)	Judgement needed (e.g., required area adjustment in epidemiological analysis if spatial extent of the study area was >10,000 km², determination of whether exposures met longterm criteria)
All pollutants ^a except PM ₁₀ , PM _{coarse} , and PM _{2.5}	Surface, satellite and personal monitoring	≤5 km; operationalized as up to 5 km between the	≤5 km	Residential address as exact address, neighborhood, census tract or block, or postal code (but not	Judgement needed (e.g., unclear monitor density, determination of whether exposures met long-term criteria)

		residence and the monitor, or up to 10 km between monitors, or at least one site per 50 km ²		city or county)	
PM ₁₀ PM _{coarse} PM _{2.5}	Surface, satellite and personal monitoring	Excluded	Excluded	Excluded	Excluded
Indirect traffic measures (Metrics based on distance or traffic density)	Objective	≤1,000 m from a highway or a major road	≤100 m	Residential address as exact address or detailed postal code (i.e., street segment)	Assumed by method

a The pollutants included were NO₂, NO_x, NO, CO, EC (including related metrics such as black carbon, black smoke, and PM absorbance), UFP, non-tailpipe PM trace metals (e.g., Cu, Fe and Zn), PAHs, benzene, PM₁₀, PM_{coarse} and PM_{2.5}.

Appendix Table 2. Search strategy for the outcome mortality.

PECOS		PubMed search terms
Population		adult[tiab] OR adults[tiab] OR child[tiab] OR children[tiab] OR pupils[tiab] OR preschooler[tiab] OR preschoolers[tiab] OR student[tiab] OR students[tiab] OR adolescent[tiab] OR adolescents[tiab] OR infant[tiab] OR infants[tiab] OR toddlers[tiab] OR newborn[tiab] OR baby[tiab] OR babies[tiab] OR person[tiab] OR persons[tiab] OR human[tiab] OR humans[tiab] OR people[tiab] OR man[tiab] OR men[tiab] OR woman[tiab] OR women[tiab] OR elderly[tiab] OR boy[tiab] OR boys[tiab] OR girl[tiab] OR girls[tiab] OR patients[tiab] OR population[tiab] OR populations[tiab] OR survivor[tiab] OR survivors[tiab] OR spouse[tiab] OR spouses[tiab] OR wife[tiab] OR husband[tiab] OR smoker[tiab] OR veteran[tiab] OR mother[tiab] OR mothers[tiab] OR father[tiab] OR fathers[tiab] OR "population based"[tiab] OR "cohort"[tiab] OR (("persons"[Mesh] OR "humans"[Mesh]))
Exposure	General terms to be combined with pollutants Different pollutants to be combined with OR	("Environmental Exposure"[Mesh] OR "Environmental Pollution"[Mesh] OR "Air Pollutants"[Mesh] OR "Air Pollution"[Itiab] OR "air pollution"[tiab] OR "air pollutants"[tiab] OR "polluted atmosphere"[tiab] OR "atmospheric pollution"[tiab] OR "polluted air"[tiab] OR "ambient air"[tiab] OR "Inhalation Exposure/adverse effects"[Mesh] OR "Motor Vehicles"[Mesh] OR "Vehicle Emissions"[Mesh] OR "traffic—related"[tiab]) OR ((traffic OR transport) AND air)
	NO _x	((("Nitrogen Oxides"[Mesh] OR "Nitrogen dioxide"[tiab] OR "NO2"[tiab] OR "NO(2)"[tiab] OR "NOx"[tiab] OR "NO(x)"[tiab] OR "Nitrogen oxide"[tiab] OR "nitrogen oxides"[tiab]))) OR "oxides of nitrogen"[tiab]
	СО	"Carbon Monoxide"[Mesh] OR "carbon monoxide"[tiab]
	Traffic PM	"Particulate Matter"[Mesh:NoExp] OR "Smog"[Mesh] OR "smog"[tiab] OR "Particle Size"[Mesh] OR "PM10"[tiab] OR PM2.5[tiab] OR PM10–2.5[tiab] OR PM2.5–10[tiab] OR PM1[tiab] OR "fine particulate"[tiab] OR "PM10"[tiab] OR "PM2.5"[tiab] OR "PM10–2.5"[tiab] OR "PM2.5–10"[tiab] OR "PM1"[tiab] OR "PM(10)"[tiab] OR "PM(2.5)"[tiab] OR "PM(10–2.5)"[tiab] OR "PM(10–2.5)"[tiab] OR "PM(10–2.5)"[tiab] OR "PM(10–2.5)"[tiab] OR

		"particulate matter"[tiab] OR "PMcoarse"[tiab] OR "PMcoarse"[tiab]
e	Non–tailpipe emissions and netals	resuspended dust[tiab] OR re—suspended dust[tiab] OR road dust[tiab] OR brake dust[tiab] OR tire dust[tiab] OR tyre dust[Text Word] OR brake wear[tiab] OR tire wear[tiab] OR tyre wear[tiab] OR road wear[tiab] OR debris dust[tiab] OR fugitive dust[tiab] OR diffuse dust[tiab] OR wear dust[tiab] OR non—exhaust[tiab] OR source apportionment[tiab] OR windblown dust[tiab] OR non—tailpipe[tiab] OR mineral dust[tiab]
		(nickel[tiab] OR Ni[tiab] OR Copper[tiab] OR Cu[tiab] OR aluminium[tiab] OR aluminium[tiab] OR Al[tiab] OR zinc[tiab] OR Zn[tiab] OR barium[tiab] OR Ba[tiab] OR iron[tiab] OR Fe[tiab] OR copper[tiab] OR Cu[tiab] OR Antimon[tiab] OR Sb[tiab] OR Tinn[tiab] OR Sn[tiab] OR Zirconium[tiab] OR Zr[tiab] OR "trace metals"[tiab]
		AND ("Particulate Matter"[Mesh:NoExp] OR "Smog"[Mesh] OR "smog"[tiab] OR "Particle Size"[Mesh] OR "PM10"[tiab] OR PM2.5[tiab] OR PM10–2.5[tiab] OR PM2.5–10[tiab] OR PM1[tiab] OR "fine particulate"[tiab] OR "PM10"[tiab] OR "PM2.5"[tiab] OR "PM10–2.5"[tiab] OR "PM2.5–10"[tiab] OR "PM1"[tiab] OR "PM(10)"[tiab] OR "PM(2.5)"[tiab] OR "PM(10–2.5)"[tiab] OR "PM(2.5–10)"[tiab] OR "PM(1)"[tiab] OR "particulate matter"[tiab] OR "PMcoarse"[tiab] OR "PMcoarse"[tiab]))
U	JFPs	"submicron"[tiab] OR "surface area"[tiab] OR "ultrafine"[tiab] OR "ultrafine particles"[tiab] OR "ultrafine particle"[tiab] OR "nano particle"[tiab] OR "nano particles"[tiab] OR "nanoparticles"[tiab] OR PM0.1[tiab] OR "PM0.1"[tiab] OR "PM0.25[tiab] OR "PM0.25"[tiab] OR "PM0.25"[tiab] OR "quasi—ultrafine"[tiab] OR "quasi ultrafine"[tiab] OR "PNC"[tiab] OR "accumulation mode"[tiab] OR "particle number"[tiab] OR "number of particles"[tiab] OR "aitken mode"[tiab]
S	Soot/BC	"Soot"[Mesh] OR soot[tiab] OR "PM2.5 absorbance"[tiab] OR "PM2.5absorbance"[tiab] OR "PM2.5abs"[tiab] OR "black carbon"[tiab] OR "carbon black"[tiab] OR "organic carbon"[tiab] OR "elemental carbon"[tiab] OR "black smoke"[tiab]
P	PAHs	"Polycyclic Aromatic Hydrocarbons"[Mesh:NoExp] OR "polycyclic aromatic hydrocarbons"[tiab] OR PAH[tiab] OR

		"PAH's"[tiab] OR PAHs[tiab] OR "benzo(a)pyrene"[tiab] OR
		benzopyrene[tiab]
	Benzene	"benzene"[Mesh] OR benzene[tiab] OR BTEX[tiab]
	Proxy measures for traffic	((((traffic[tiab]) NOT ("Accidents, Traffic"[Mesh] OR safety[tiab] OR accident[tiab] OR accidents[tiab] OR injur*[tiab] OR collision*[tiab] OR crash*[tiab])) OR "traffic intensity"[tiab] OR "traffic density"[tiab] OR "traffic load"[tiab] OR "traffic count"[tiab] OR "road length"[tiab] OR ((proximity[tiab] OR near[tiab] OR distance[tiab] OR nearest[tiab] OR next[tiab] OR close[tiab] OR closest[tiab]) AND (road*[tiab] OR highway*[tiab] OR freeway*[tiab] OR motorway*[tiab] OR interstate[tiab] OR expressway[tiab])))) OR ((vehicle[tiab] OR vehicles[tiab] OR vehicular[tiab] OR auto[tiab] OR automobile[tiab] OR bus[tiab] OR buses[tiab] OR car[tiab] OR truck[tiab] OR trucker[tiab] OR trucks[tiab] OR engine[tiab] OR transport[tiab] OR traffic[tiab]) AND (emissions[tiab] OR exhaust[tiab] OR fume*[tiab]))
Comparator	Measures of effect	"risk" [Mesh] OR "risk" [tiab] OR "risks" [tiab] OR "incidence" [Mesh] OR "incidence" [tiab] OR "incident" [tiab] OR "Prevalence" [Mesh] OR "prevalence" [tiab] OR "prevalent" [tiab] OR "Risk Factors" [Mesh] OR "risk factor" [tiab] OR "Odds Ratio" [Mesh] OR "odds" [tiab] OR "onset" [tiab] OR "associated" [tiab] OR "association" [tiab] OR "cause" [tiab] OR "causes" [tiab] OR "caused" [tiab] OR "develop" [tiab] OR "developed" [tiab] OR "prevent" [tiab] OR "prevents" [tiab] OR "prevented" [tiab] OR "increase" [tiab] OR "increased" [tiab] OR "increases" [tiab] OR "effects" [tiab] OR "effects" [tiab] OR "affect" [tiab] OR "affects" [tiab] OR "affected" [tiab] OR "protective" [tiab] OR "protect" [tiab] OR "protected" [tiab] OR "harm" [tiab] OR "harms" [tiab] OR "harmed" [tiab] OR "harmful" [tiab] OR "hazard" [tiab] OR "proportional hazard" [tiab]
Outcome	Mortality	("Mortality"[Mesh] OR "mortality"[Mesh] OR "Myocardial Ischemia/mortality"[Mesh] OR "Respiratory Tract Diseases/mortality"[Mesh] OR "Respiratory Tract Diseases/mortality"[Mesh] OR "Respiratory Tract Infections/mortality"[Mesh] OR "Respiration Disorders/mortality"[Mesh] OR "Lung Neoplasms/mortality"[Mesh] OR "Pulmonary Disease, Chronic Obstructive/mortality"[Mesh]) OR (("cause—specific"[tiab] OR "all—cause"[tiab] OR "non—accidental"[tiab] OR "natural"[tiab] OR "natural—cause"[tiab] OR "cardiovascular"[tiab] OR "respiratory"[tiab] OR "cardio

		respiratory"[tiab] OR "lung cancer"[tiab] OR "COPD"[tiab]) AND (mortality[tiab] OR death[tiab] OR "deadly"[tiab] OR died[tiab] OR fatal*[tiab] OR surviv*[tiab])) OR ("mortality"[tiab] OR "death"[tiab])	
Study	Filters	NOT ((((((("shortterm"[ti] OR "short—term"[ti] OR "time series"[ti] OR time—series[ti]) AND (("shortterm"[ti] OR "short—term"[ti] OR "time series"[ti] OR time—series[ti]) NOT ("longterm"[tiab] OR "long term"[tiab] OR "medium term"[tiab] OR "intermediate term"[tiab] OR "chronic"[tiab]))))) OR ("Clinical Trial"[Publication Type] OR "Treatment Outcome"[MeSH] OR "Cross—Over Studies"[Mesh] OR "case cross over"[tiab])) OR ("Air Pollutants, Occupational"[Mesh] OR "Accidents, Traffic"[Mesh] OR "Protective Devices"[Mesh])) OR (mouse[Title/Abstract] OR mice[Title/Abstract] OR rat[Title/Abstract] OR rats[Title/Abstract]) AND English[Language] AND ("1980/01/01"[Datea — Publication] : "3000"[Datea — Publication])	
	LUDOK	(Sterblichkeit[methods] AND 7L) OR (road[title] AND 7L) OR (traffic[title] AND 7L) OR (schwangerschaft[title] AND 7L) OR (geburt[methods] AND 7L) OR (arteriosklerose[methods] AND 7L) OR (diabetes[methods] AND 7L) OR (leukämie[methods] AND 7L) OR (4O AND 7L) OR (4I AND 7L)	

^a Pubmed was searched on Feb 12, 2019 (11,461 records identified) and updated on July 31, 2019 (657 additional records).

^b Ludok was searched on October, 21 2019 (3,935 records). 7L is the code for long-term studies, 4O code for outcomes related to pregnancy and prenatal development, 4I outcomes related to outcomes regarding neurocognitive outcomes, 4B = lung function, 4E = acute respiratory outcomes, 4H = cardiovascular outcomes like stroke, blood pressure, 4F = chronic respiratory outcomes, [] indicates the field of searched in the database. the [methods]-field is where LUDOK saves the keywords.

Appendix Table 3. List of excluded mortality studies with reason (N = 119).

	Reference	Exclusion rationale
	Abbey, 1993, Chronic disease associated with long-term	No quantitative measure of
1.	concentrations of nitrogen dioxide	association
	Abbey, 1995, Long-term ambient concentrations of particulates	PM monitoring or satellite data
	and oxidants and development of chronic disease in a cohort of	
2.	nonsmoking California residents	
	Abbey, 1999, Long-term inhalable particles and other air	Spatial scale
3.	pollutants related to mortality in nonsmokers	
	Andersen, 2015, A study of the combined effects of physical	No quantitative measure of
	activity and air pollution on mortality in elderly urban residents:	association
4.	The Danish Diet, Cancer, and Health cohort	
	Bateson, 2004, Who is sensitive to the effects of particulate air	Study design
	pollution on mortality? A case-crossover analysis of effect	
5.	modifiers	
	Baxter, 2013, Examining the effects of air pollution composition	Review, methodological, HIA, or
6.	on within region differences in PM _{2.5} mortality risk estimates	similar paper (no primary data)
	Bentayeb, 2015, Association between long-term exposure to air	Nationwide/statewide study with no
	pollution and mortality in France: A 25-year follow-up study	or insufficient area-specific
7.		adjustments
	Beverland, 2012, A comparison of short-term and long-term air	Review, methodological, HIA, or
	pollution exposure associations with mortality in two cohorts in	similar paper (no primary data)
8.	Scotland	
	Bidoli, 2016, Residential proximity to major roadways and lung	Study design
9.	cancer mortality. Italy, 1990-2010: An observational study	
	Blount, 2017, Traffic-related air pollution and all-cause mortality	Very selective subgroup
10.	during tuberculosis treatment in California	
	Burstyn, 2005, Polycyclic aromatic hydrocarbons and fatal	Occupational study
11.	ischemic heart disease	
	Cakmak, 2016, Ozone exposure and cardiovascular-related	PM monitoring or satellite data
	mortality in the Canadian Census Health and Environment Cohort	
12.	(CanCHEC) by spatial synoptic classification zone	
	Cakmak, 2018, Associations between long-term pm and ozone	Spatial scale
	exposure and mortality in the Canadian Census Health and	
1.0	Environment Cohort (CanCHEC), by spatial synoptic classification	
13.	zone	
	Cao, 2011, Association between long-term exposure to outdoor	Spatial scale
14.	air pollution and mortality in China: A cohort study	
	Cesaroni, 2012, Nitrogen dioxide levels estimated from land use	Review, methodological, HIA, or
4.5	regression models several years apart and association with	similar paper (no primary data)
15.	mortality in a large cohort study	Custial and
1.0	Chen, 2005, The association between fatal coronary heart disease	Spatial scale
16.	and ambient particulate air pollution: Are females at greater risk?	DNA manifesting and a little of the
47	Chen, 2016, Ambient fine particulate matter and mortality among	PM monitoring or satellite data
17.	survivors of myocardial infarction: Population-based cohort study	

	Reference	Exclusion rationale
18.	Chen, 2019, Long-term exposure to air pollution and survival after ischemic stroke	Spatial scale
19.	Cohen, 2017, Long-term exposure to traffic-related air pollution and cancer among survivors of myocardial infarction: A 20-year follow-up study	Health outcome
20.	Crouse, 2015, Ambient PM _{2.5} , O ₃ , and NO₂ exposures and associations with mortality over 16 years of follow-up in the Canadian Census Health and Environment Cohort (CanCHEC)	Nationwide/statewide study with no or insufficient area-specific adjustments
20.	Crouse, 2016, A new method to jointly estimate the mortality risk of long-term exposure to fine particulate matter and its components	Review, methodological, HIA, or similar paper (no primary data)
22.	Crouse, 2019, Complex relationships between greenness, air pollution, and mortality in a population-based Canadian cohort.	PM monitoring or satellite data
23.	Dehbi, 2017, Air pollution and cardiovascular mortality with over 25 years follow-up: a combined analysis of two British cohorts	Nationwide/statewide study with no or insufficient area-specific adjustments
24.	Di, 2017, Air pollution and mortality in the Medicare population	Spatial scale
25.	Dockery, 1993, An association between air pollution and mortality in six US cities	No within-area or spatial contrast exploited
26.	Domínguez-Berjón, 2016, Lung cancer and urbanization level in a region of Southern Europe: Influence of socio-economic and environmental factors	Study design
27.	Dominici, 2005, Revised analyses of the National Morbidity, Mortality, and Air Pollution Study: Mortality among residents of 90 cities	Study design
28.	Dong, 2012, Long-term exposure to ambient air pollution and respiratory disease mortality in Shenyang, China: A 12-year population-based retrospective cohort study	No within-area or spatial contrast exploited
29.	Eckel, 2016, Air pollution affects lung cancer survival	Very selective subgroup
30.	Elliott, 2007, Long-term associations of outdoor air pollution with mortality in Great Britain	Study design
31.	Eum, 2019, Long-term NO exposures and cause-specific mortality in American older adults.	Spatial scale
32.	Filleul, 2005, Twenty five year mortality and air pollution: Results from the French PAARC survey	Nationwide/statewide study with no or insufficient area-specific adjustments
33.	Fischer, 2015, Air pollution and mortality in seven million adults: The Dutch Environmental Longitudinal Study (DUELS)	Nationwide/statewide study with no or insufficient area-specific adjustments
34.	Garcia, 2016, Association of long-term PM _{2.5} exposure with mortality using different air pollution exposure models: impacts in rural and urban California	PM monitoring or satellite data
35.	Greven, 2011, An approach to the estimation of chronic air pollution effects using spatio-temporal information	Review, methodological, HIA, or similar paper (no primary data)
36.	Habermann, 2012, Motor vehicle traffic and cardiovascular mortality in male adults	Spatial scale

	Reference	Exclusion rationale
37.	Hales, 2012, Air pollution and mortality in New Zealand: Cohort study	Nationwide/statewide study with no or insufficient area-specific adjustments
38.	Halonen, 2015, Road traffic noise is associated with increased cardiovascular morbidity and mortality and all-cause mortality in London	No quantitative measure of association
39.	Halonen, 2016, Is long-term exposure to traffic pollution associated with mortality? A small-area study in London	Study design
40.	Hart, 2011, Long-term ambient multipollutant exposures and mortality	Nationwide/statewide study with no or insufficient area-specific adjustments
41.	Hart, 2015, The association of long-term exposure to PM _{2.5} on all-cause mortality in the Nurses' Health Study and the impact of measurement-error correction	Nationwide/statewide study with no or insufficient area-specific adjustments
42.	Hartiala, 2016, Ambient air pollution is associated with the severity of coronary atherosclerosis and incident myocardial infarction in patients undergoing elective cardiac evaluation	Spatial scale
43.	Hayes, 2019, PM _{2.5} air pollution and cause-specific cardiovascular disease mortality	Nationwide/statewide study with no or insufficient area-specific adjustments
44.	Héritier, 2018, A systematic analysis of mutual effects of transportation noise and air pollution exposure on myocardial infarction mortality: A nationwide cohort study in Switzerland	Nationwide/statewide study with no or insufficient area-specific adjustments
45.	Hoek, 2002, Association between mortality and indicators of traffic-related air pollution in the Netherlands: A cohort study	Pilot study; complete results in Beelen, 2008
46.	Iwai, 2005, Correlation between suspended particles in the environmental air and causes of disease among inhabitants: cross-sectional studies using the vital statistics and air pollution data in Japan	No within-area or spatial contrast exploited
47.	Jerrett, 2005, Spatial analysis of air pollution and mortality in Los Angeles	Spatial scale
48.	Jerrett, 2009, Long-term ozone exposure and mortality	No within-area or spatial contrast exploited
49.	Jerrett, 2013, Spatial analysis of air pollution and mortality in California	Nationwide/statewide study with no or insufficient area-specific adjustments
50.	Katanoda, 2011, An association between long-term exposure to ambient air pollution and mortality from lung cancer and respiratory diseases in Japan	No within-area or spatial contrast exploited
51.	Kim, 2017, Association between long-term exposure to particulate matter air pollution and mortality in a South Korean national cohort: comparison across different exposure assessment approaches	Nationwide/statewide study with no or insufficient area-specific adjustments
52.	Kim, 2017, Cardiovascular effects of long-term exposure to air pollution: A population-based study with 900,845 person-years of follow-up	Analytical error

	Reference	Exclusion rationale
53.	Knox, 1997, Hazard proximities of childhood cancers in Great Britain from 1953-80	Study design
54.	Li, 2018, All-cause mortality risk associated with long-term exposure to ambient PM in China: A cohort study.	PM monitoring or satellite data
55.	Liang, 2018, Satellite-based short- and long-term exposure to pm and adult mortality in urban Beijing, China	PM monitoring or satellite data
56.	Lim, 2018, Association between long-term exposure to ambient air pollution and diabetes mortality in the US	Nationwide/statewide study with no or insufficient area-specific adjustments
57.	Lim, 2019, Long-term exposure to ozone and cause-specific mortality risk in the US	Nationwide/statewide study with no or insufficient area-specific adjustments
58.	Lim, 2019, Mediterranean diet and the association between air pollution and cardiovascular disease mortality risk	Nationwide/statewide study with no or insufficient area-specific adjustments
59.	Lipfert, 2002, Temporal and spatial relations between age specific mortality and ambient air quality in the United States: regression results for counties, 1960-97	Study design
60.	Lipfert, 2006, PM _{2.5} constituents and related air quality variables as predictors of survival in a cohort of US military veterans	Nationwide/statewide study with no or insufficient area-specific adjustments
61.	Lipfert, 2006, Traffic density as a surrogate measure of environmental exposures in studies of air pollution health effects: long-term mortality in a cohort of US veterans	Spatial scale
62.	Lipfert, 2008, Vehicular traffic effects on survival within the Washington university-EPRI veterans cohort: New estimates and sensitivity studies	Spatial scale
63.	Lipfert, 2009, Air pollution and survival within the Washington university-EPRI veterans cohort: Risks based on modeled estimates of ambient levels of hazardous and criteria air pollutants	Spatial scale
64.	Lipsett, 2011, Long-term exposure to air pollution and cardiorespiratory disease in the California Teachers Study cohort	Nationwide/statewide study with no or insufficient area-specific adjustments
65.	Liu, 2008, Ambient exposure to criteria air pollutants and female lung cancer in Taiwan	Spatial scale
66.	Loop, 2018, Fine particulate matter and incident coronary heart disease in the REGARDS cohort	Spatial scale
67.	Maheswaran, 2003, Stroke mortality associated with living near main roads in England and Wales: A geographical study	Study design
68.	Maheswaran, 2006, Outdoor NO _x and stroke mortality: adjusting for small area level smoking prevalence using a Bayesian approach	Study design
69.	Makar, 2017, Estimating the causal effect of low levels of fine particulate matter on death and hospitalization: are levels below the safety standards harmful?	Nationwide/statewide study with no or insufficient area-specific adjustments

	Reference	Exclusion rationale
	Malik, 2019, Association of long-term exposure to particulate	Nationwide/statewide study with no
	matter and ozone with health status and mortality in patients	or insufficient area-specific
70.	after myocardial infarction.	adjustments
	McDonnell, 2000, Relationships of mortality with the fine and	Spatial scale
	coarse fractions of long-term ambient PM ₁₀ concentrations in	
71.	nonsmokers	
	Miller, 2007, Long-term exposure to air pollution and incidence of	PM monitoring or satellite data
72.	cardiovascular events in women	_
	Nawrot, 2011, The impact of traffic air pollution on bronchiolitis	Very selective subgroup
73.	obliterans syndrome and mortality after lung transplantation	
	Pinault, 2016, Risk estimates of mortality attributed to low	PM monitoring or satellite data
	concentrations of ambient fine particulate matter in the Canadian	
74.	Community Health Survey cohort	
	Pinault, 2017, Associations between fine particulate matter and	PM monitoring or satellite data
	mortality in the 2001 Canadian Census Health and Environment	
75.	Cohort	
	Pinault, 2018, Diabetes status and susceptibility to the effects of	PM monitoring or satellite data
	PM _{2.5} exposure on cardiovascular mortality in a National	
76.	Canadian Cohort	
	Pope, 2002, Lung cancer, cardiopulmonary mortality, and long-	No within-area or spatial contrast
77.	term exposure to fine particulate air pollution	exploited
	Pope, 2009, Cardiovascular mortality and exposure to airborne	PM monitoring or satellite data
	fine particulate matter and cigarette smoke: Shape of the	
78.	exposure-response relationship	
	Pope, 2015, Relationships between fine particulate air pollution,	Nationwide/statewide study with no
	cardiometabolic disorders, and cardiovascular mortality	or insufficient area-specific
79.		adjustments
	Pope, 2019, Mortality risk and fine particulate air pollution in a	PM monitoring or satellite data
80.	large, representative cohort of US adults	
	Puett, 2009, Chronic fine and coarse particulate exposure,	Nationwide/statewide study with no
04	mortality, and coronary heart disease in the Nurses' Health Study	or insufficient area-specific
81.	Duett 2011 Destinulate meetter over according and the con-	Alationwide (statewide study with no
	Puett, 2011, Particulate matter exposures, mortality, and	Nationwide/statewide study with no
റാ	cardiovascular disease in the Health Professionals follow-up study	or insufficient area-specific
82.	Dun 2017 Long torm DM ovnosure and respiratory series and	adjustments
ດາ	Pun, 2017, Long-term PM _{2.5} exposure and respiratory, cancer, and	PM monitoring or satellite data
83.	cardiovascular mortality in older US adults Pihoiro 2018, Incidence and mortality rick for recoiratory tract	Study docign
	Ribeiro, 2018, Incidence and mortality risk for respiratory tract	Study design
0.4	cancer in the city of São Paulo, Brazil: Bayesian analysis of the association with traffic density	
84.	,	Study design
OE	Ribeiro, 2019, Incidence and mortality for respiratory cancer and	Study design
85.	traffic-related air pollution in São Paulo, Brazil	Study docign
96	Rodrigues, 2017, Risk factors in cardiovascular disease mortality	Study design
86.	associated with high exposure to vehicular traffic	

	Reference	Exclusion rationale
87.	Ruttens, 2017, An association of particulate air pollution and traffic exposure with mortality after lung transplantation in Europe	Very selective subgroup
88.	Samoli, 2003, Investigating the dose-response relation between air pollution and total mortality in the APHEA-2 multicity project	Study design
89.	Samoli, 2005, Estimating the exposure-response relationships between particulate matter and mortality within the APHEA multicity project	Study design
90.	Sanyal, 2018, Long-term effect of outdoor air pollution on mortality and morbidity: A 12-year follow-up study for metropolitan France.	Study design
91.	Schwartz, 2018, Estimating the effects of PM on life expectancy using causal modeling methods	Spatial scale
92.	Scoggins, 2004, Spatial analysis of annual air pollution exposure and mortality	Study design
93.	Silveira, 2018, Green spaces and mortality due to cardiovascular diseases in the city of Rio de Janeiro	Study design
94.	Son, 2011, Survival analysis of long-term exposure to different sizes of airborne particulate matter and risk of infant mortality using a birth cohort in Seoul, Korea	PM monitoring or satellite data
95.	Strak, 2017, Long-term exposure to particulate matter, NO ₂ and the oxidative potential of particulates and diabetes prevalence in a large national health survey	Nationwide/statewide study with no or insufficient area-specific adjustments
96.	Sunyer, 2001, Particles, and not gases, are associated with the risk of death in patients with chronic obstructive pulmonary disease	Study design
97.	Theophanides, 2007, Mortality and pollution in several Greek cities	Study design
98.	Thurston, 2016, Ambient particulate matter air pollution exposure and mortality in the NIH-AARP Diet and Health Cohort	Spatial scale
99.	Thurston, 2016, Ischemic heart disease mortality and long-term exposure to source-related components of US fine particle air pollution	Spatial scale
100.	Tonne, 2008, Air pollution and mortality benefits of the London congestion charge: Spatial and socioeconomic inequalities	Review, methodological, HIA, or similar paper (no primary data)
101.	Tran, 2018, Impact of air pollution on cause-specific mortality in Korea: results from Bayesian model averaging and principal component regression approaches	No quantitative measure of association
102.	Turner, 2014, Interactions between cigarette smoking and fine particulate matter in the Risk of Lung Cancer Mortality in Cancer Prevention Study II	Nationwide/statewide study with no or insufficient area-specific adjustments
103.	Turner, 2016, Long-term ozone exposure and mortality in a large prospective study	Nationwide/statewide study with no or insufficient area-specific adjustments
104.	Turner, 2017, Interactions between cigarette smoking and ambient PM for cardiovascular mortality	Spatial scale

	Reference	Exclusion rationale
	Vandentorren, 2003, Long-term mortality among adults with or	No relevant exposure metric
105.	without asthma in the PAARC study	
	Vanos, 2013, Synoptic weather typing applied to air pollution	Study design
106.	mortality among the elderly in 10 Canadian cities	
	Villeneuve, 2015, Long-term exposure to fine particulate matter	Spatial scale
107.	air pollution and mortality among Canadian women	
	Vineis, 2007, Lung cancers attributable to environmental tobacco	Health outcome
	smoke and air pollution in non-smokers in different European	
108.	countries: A prospective study	
100	Wang, 2009, Long-term exposure to gaseous air pollutants and	Study design
109.	cardio-respiratory mortality in Brisbane, Australia	224
	Weichenthal, 2016, Oxidative burden of fine particulate air	PM monitoring or satellite data
110	pollution and risk of cause-specific mortality in the Canadian	
110.	Census Health and Environment Cohort (CanCHEC) Weichenthal, 2017, Impact of oxidant gases on the relationship	Nationwide (statewide study with no
	between outdoor fine particulate air pollution and nonaccidental,	Nationwide/statewide study with no or insufficient area-specific
111.	cardiovascular, and respiratory mortality	adjustments
111.	Wen, 2012, Air pollution shortens life expectancy and health	No relevant exposure metric
112.	expectancy for older adults: The case of China	No relevant exposure metric
	Wong, 2015, Satellite-based estimates of long-term exposure to	PM monitoring or satellite data
	fine particles and association with mortality in elderly Hong Kong	a continue data
113.	residents	
	Wong, 2016, Cancer mortality risks from long-term exposure to	PM monitoring or satellite data
114.	ambient fine particle	
	Woodruff, 2006, Fine particulate matter (PM _{2.5}) air pollution and	PM monitoring or satellite data
115.	selected causes of postneonatal infant mortality in California	
	Woodruff, 2008, Air pollution and postneonatal infant mortality	Spatial scale
116.	in the United States, 1999-2002	
	Xu, 2013, Health effects of air pollution on length of respiratory	PM monitoring or satellite data
117.	cancer survival	
	Zhang, 2011, Long-term exposure to ambient air pollution and	Analytical error
	mortality due to cardiovascular disease and cerebrovascular	
118.	disease in Shenyang, China	
	Zúñiga, 2016, Assessment of the possible association of air	No within-area or spatial contrast
	pollutants PM ₁₀ , O ₃ , NO ₂ with an increase in cardiovascular,	exploited
110	respiratory, and diabetes mortality in Panama City: A 2003 to	
119.	2013 data analysis	

Appendix Table 4. Evaluation of impact of extracted effect estimate for NO_2 , $PM_{2.5}$, and $EC^{a.}$

	Extracted effect estimate	Authors- favored estimate	Fuly adjusted estimate	Reason for deviation or comment	Increment
NO ₂					
Beelen 2014	1.01 (0.99, 1.03)	Same	Same		10 μg/m ³
Beelen 2008	1.08 (1.00, 1.16)	Same	Same		30 μg/m ³
Carey 2013	1.02 (1.00, 1.04)	1.03 (1.00, 1.05)	1.03 (1.00, 1.05)	All adjusted for age, sex, smoking, BMI, income, but we prefered the estimate with an area correction (Table E7). Also RRs reported for different SES indicators corrections instead of income (no area level correction): Employment: 1.04 (1.01, 1.07) Education: 1.06 (1.03, 1.08) Those estimates are slighly larger than the	10.7 μg/m³
		_	_	one we extracted.	4.2
Cesaroni 2013	1.03 (1.02, 1.03)	Same	Same		10 μg/m³
Crouse 2015	1.05 (1.04, 1.07)	1.05 (1.03, 1.07)	1.05 (1.03, 1.07)	All adjusted for individual and contextual covariates with city as a radom effect, indirectly adjusting for smoking and BMI, but we prefered the models particioned into within-city contrast as oposed to the overall contrast.	5 ppb
Dirgawati 2019	1.06 (1.00, 1.13)	Same	Same	RR for average exposure period per protocol. Also RRs reported for baseline 1.06 (1.00, 1.13) and past-year exposure 1.07 (1.01, 1.13).	10 μg/m ³

Hanigan 2019	1.03 (0.98, 1.07)	Same	Same		5 μg/m³
Hvidtfeldt 2019	1.07 (1.04, 1.10)	Same	Same		10 μg/m³
Nieuwenhuijsen 2018	1.01 (1.00, 1.02)	Same	Same		5 μg/m³
Yang 2018	1 00 (0 07 1 02)	Same	Same		25.6 μg/m³
	1.00 (0.97, 1.03)		Same		10 μg/m ³
Yorifuji 2013	1.12 (1.07, 1.18)	Same	Same		10 μg/m
EC Paplan 2014	1 02 (0 07 1 07)	Corres	Come		1 1 1 1 0 - 5 /
Beelen 2014	1.02 (0.97, 1.07)	Same	Same		1 1x10 ⁻⁵ /m
Beelen 2008	1.05 (1.00, 1.11)	Same	Same		10 μg/m ³
Dirgawati 2019	1.12 (1.02, 1.22)	Same	Same	RR for average exposure period per protocol.	1 1x10 ⁻⁵ /m
				Also RRs reported for baseline 1.12 (1.02, 1.23) and past-year exposure 1.14 (1.03, 1.24).	
Hansell 2016	1.02 (1.01, 1.04)	Same	Same	RR for 1971 baseline exposure, and for most recent outcome data (2002-2009). We picked the most recent years because we also extracted the PM ₁₀ estimate, which was only available for 2002-2009. Also RRs reported for other years: 1972-2009: 1.03 (1.02, 1.05) 1972-1981: 1.05 (1.02, 1.08) 1982-1991: 1.03 (1.01, 1.06) 1992-2001: 1.04 (1.02, 1.05)	10 μg/m³
				Those estimates are slighly larger than the one we extracted. Also effect estimates are reported for more recent exposures (1981 and 1991). Those were also larger than what we extracted.	

Hvidtfeldt 2019	1.09 (1.04, 1.15)	Same	Same		1 μg/m³
Nieuwenhuijsen 2018	1.02 (1.00, 1.05)	Same	Same		1 1x10 ⁻⁵ /m
Yang 2018	1.03 (1.00, 1.05)	Same	Same		9.6 μg/m ³
Badaloni 2017	1.03 (1.01, 1.05)	Same	Same		1.5 1x10 ⁻⁵ /m
Yap 2012, Renfrew/Paisley	1.08 (1.02, 1.15)	1.10 (1.04, 1.17)	1.08 (1.02, 1.15)	RRs from the most detailed exposure model (MultiBS). All estimates were corrected for marital status, body mass index, smoking, cholesterol, systolic blood pressure, and social class. We extracted the fully adjusted model, which was additionally adjusted for	10 μg/m³
Yap 2012. Collaborative cohorts	1.01 (0.95, 1.06)	1.01 (0.96, 1.05)	1.01 (0.95, 1.06)	area-level deprivation category. RRs from the most detailed exposure model (MultiBS). All estimates were corrected for marital status, body mass index, smoking, cholesterol, systolic blood pressure, and social class. We extracted th efully adjusted model, which was additionally adjusted for area-level deprivation category.	10 μg/m³
Ostro 2015	1.00 (0.97, 1.04)	Same	Same		0.8 μg/m ³
PM _{2.5}	,				
Badaloni 2017	1.03 (1.01, 1.05)	Same	Same		6.6 μg/m ³
Beelen 2014	1.07 (1.02, 1.13)	Same	Same		5 μg/m³
Beelen 2008	1.06 (0.97, 1.16)	Same	Same		10 μg/m³

Carey 2013	1.00 (0.98, 1.02)	1.02 (1.00, 1.05)	1.02 (1.00, 1.05)	All adjusted for age, sex, smoking, BMI, income, but we prefered the estimate with an area correction (Table E7). Also RRs reported for different SES indicators corrections instead of income (no area level correction): Employment: 1.04 (1.02, 1.07) Education: 1.04 (1.02, 1.06)	1.9 μg/m³
				extracted.	
Dirgawati 2019	1.07 (0.98, 1.16)	Same	Same	RR for average exposure period per protocol.	5 μg/m³
				Also RRs reported for baseline 1.06 (0.98, 1.15) and past-year exposure 1.06 (0.98, 1.16).	
Hanigan 2019	1.05 (0.98, 1.12)	Same	Same		1 μg/m³
Hvidtfeldt 2019	1.13 (1.05, 1.21)	Same	Same		5 μg/m ³
Krewski 2009, LA	1.14 (1.03, 1.27)	Same	1.12 (1.003, 1.239)	We extracted the estimate corrected for 44 individual-level covariates. The fully adjusted model additionally corrected for 4 ecologic covariates (Table 23).	10 μg/m³
Krewski 2009, NYC	0.98 (0.95, 1.02)	Same	0.98 (0.93, 1.03)	We extracted the estimate corrected for 44 individual-level covariates. The fully adjusted model additionally corrected for 7 ecologic covariates (Table 15).	1.5 μg/m³
Nieuwenhuijsen 2018	1.03 (0.99, 1.06)	Same	Same		5 μg/m ³
Ostro 2015	1.01 (0.98, 1.05)	Same	Same		9.6 μg/m ³
Yang 2018	1.03 (1.01, 1.06)	Same	Same		5.5 μg/m ³

^a It was not feasible to go back to all individual papers and extract multiple effect estimates from studies. We went back to all individual papers on NO₂, PM_{2.5} and EC and assessed the robustness of the effect estimates. We noted that in virtually all cases the extracted effect estimate was the same as the fully adjusted effect estimate.

For NO_2 in two of the 11 studies the extracted effect estimate differed from the fully adjusted model because we required an additional area adjustment for traffic-specificity reasons. In one study, the extracted effect estimate was slightly lower (Carey et al. 2013) and the other study was the same than the authors-favoured estimate, albeit with a slightly different CI) (Crouse et al. 2015).

For EC, the only difference was in the Yap et al. 2012 study, where we extracted the slightly lower fully adjusted estimate rather than the authors-favored estimate. For PM_{2.5}, we extracted a null finding instead of the positive authors-favored estimate because we required an additional area adjustment for traffic-specificity reasons (Carey et al. 2013). Furthermore, we selected the slightly higher estimate for Krewski et al. 2009 in the LA population.

We also did not find evidence that authors reported the highest effect estimates in their abstract. Mostly they reported the fully adjusted ones. In some studies, multiple exposure periods were evaluated (e.g., Hansell et al. 2016, Dirgawati et al. 2019). In these instances the lowest effect estimate was actually extracted because they reflected the most recent outcome data, for which years also pollutant data was available (Hansell et al. 2016) or the average exposure (Dirgawati et al. 2019).

Appendix Table 5. Noise-adjusted analyses of non-accidental mortality.

Reference	Study Name	Pollutant	Increment	Single pollutant	Noise-adjusted
Hvidtfeldt 2019	DDCH	NO ₂	10 μg/m³	1.07 (1.04, 1.10)	1.05 (1.01, 1.09)
		BC	1 μg/m³	1.09 (1.04, 1.15)	1.04 (0.98, 1.11)
		PM ₁₀	10 μg/m³	1.12 (1.03, 1.22)	1.03 (0.94, 1.14)
		PM _{2.5}	5 μg/m³	1.13 (1.05, 1.21)	1.06 (0.98, 1.15)
Nieuwenhuijsen 2018	Barcelona	NO ₂	5 μg/m³	1.01 (1.00, 1.02)	1.01 (1.00, 1.02)
	Mega Cohort				
		PM _{2.5 abs}	1 1x10 ⁻⁵ /m	1.02 (1.00, 1.05)	1.03 (1.00, 1.06)
		PM _{2.5}	5 μg/m³	1.03 (0.99, 1.06)	1.03 (0.99, 1.06)
Raaschou-Nielsen	DDCH	NO ₂	1 μg/m³	1.08 (0.98, 1.18)	1.08 (0.98, 1.18)
2012					
		Density	1 vehicle-km/day	1.01 (0.99, 1.03)	1.01 (0.99, 1.03)
		Distance	<50 vs. >50 m	0.94 (0.85, 1.05)	0.94 (0.85, 1.05)
Tonne 2016 ^a	London MI	NO ₂	8 μg/m³	1.04 (0.99, 1.10)	1.04 (0.97, 1.10)
	Cohort				
		NO _x	19.2 μg/m ³	1.03 (0.98, 1.08)	1.02 (0.97, 1.08)
		traffic PM _{2.5}	0.3 μg/m ³	1.02 (0.98, 1.07)	1.02 (0.97, 1.06)
		nontailpipe	0.3 μg/m ³	1.04 (1.00, 1.09)	1.04 (0.99, 1.09)
		PM _{2.5}			

^a Indicates a patient population.

Appendix Tabe 6. Risk of bias assessment for studies included in meta-analysis: non-accidental mortality.

Reference	Study Name	Pollutants in meta-	Confounding	Selection	Exposure	Outcome	Missing	Selective
		analysis		Bias	Assessment	Measurement	Data	Reporting
Badaloni 2017	Rome Longitudinal	PM _{2.5 abs} , PM ₁₀ , PM _{2.5} ,	High	Low	Mod	Low	Low	Low
		Cu, Fe						
Bauleo 2019	Civitavecchia Study	NO _x	High	Low	Mod	Low	Low	Low
Beelen 2008	NLCS-AIR	NO ₂ , BC, PM _{2.5}	Mod	Low	Mod	Low	Low	Low
Beelen 2014	ESCAPE	NO ₂ , NO _x , PM _{2.5 abs} ,	Low	Low	Mod	Low	Low	Low
		PM ₁₀ , PM _{2.5}						
Beelen 2015	ESCAPE	Cu, Fe	Low	Low	Mod	Low	Low	Low
Carey 2013	English National	NO ₂ , PM ₁₀ , PM _{2.5}	Mod	Mod	Low	Low	Low	Low
	Cohort							
Cesaroni 2013	Rome Longitudinal	NO ₂	High	Low	Low	Low	Low	Low
Crouse 2015	1991 CanCHEC	NO ₂	Mod	Low	Mod	Low	Low	Low
Dirgawati 2019	HIMS	NO ₂ , NO _x , PM _{2.5 abs} , PM _{2.5}	Low	Low	Mod	Low	Low	Low
Hanigan 2019	45 and Up Study	NO ₂ , PM _{2.5}	Low	Low	Mod	Low	Low	Low
Hansell 2016	ONS-Longitudinal	BS, PM ₁₀	High	Low	Low	Low	High	Low
Hvidtfeldt 2019	DDCH	NO ₂ , BC, PM ₁₀ , PM _{2.5}	Low	Low	Low	Low	Low	Low
Krewski 2009	ACS-CPS II LA	PM _{2.5}	Low	Low	Mod	Low	Low	Low
	ACS-CPS II NYC	PM _{2.5}	Low	Low	Mod	Low	Low	Low
Nafstad 2004	Oslo men's cohort	NO _x	Mod	Low	Low	Low	Low	Low
Nieuwenhuijsen	Barcelona Mega	NO ₂ , PM _{2.5 abs} , PM ₁₀ ,	High	Low	Low	Low	Low	Low
2018	Cohort	PM _{2.5}						
Ostro 2015	California Teachers	EC, PM _{2.5} , Cu, Fe	Low	Low	Low	Low	Low	Low
	Study							
Stockfelt 2015	PPS	NO _x	Low	Low	Low	Low	Low	Low
Yang 2018	Hong Kong Elderly	NO ₂ , BC, PM _{2.5}	Low	Mod	Mod	Low	Low	Low
Yap 2012	Renfrew/Paisley	BS	Low	Low	High	Low	Low	Low
	Collaborative cohorts	BS	Low	Low	High	Low	Low	Low
Yorifuji 2013	Shizuoka Elderly	NO ₂	Low	High	Low	Low	High	Low

Mod = moderate. ^a References have the same risk of bias ratings for all pollutants within each study.

Appendix Table 7. Risk of bias ratings and rationales for studies included in meta-analysis: non-accidental mortality.

We refer to a seperate Excel spreadsheet.

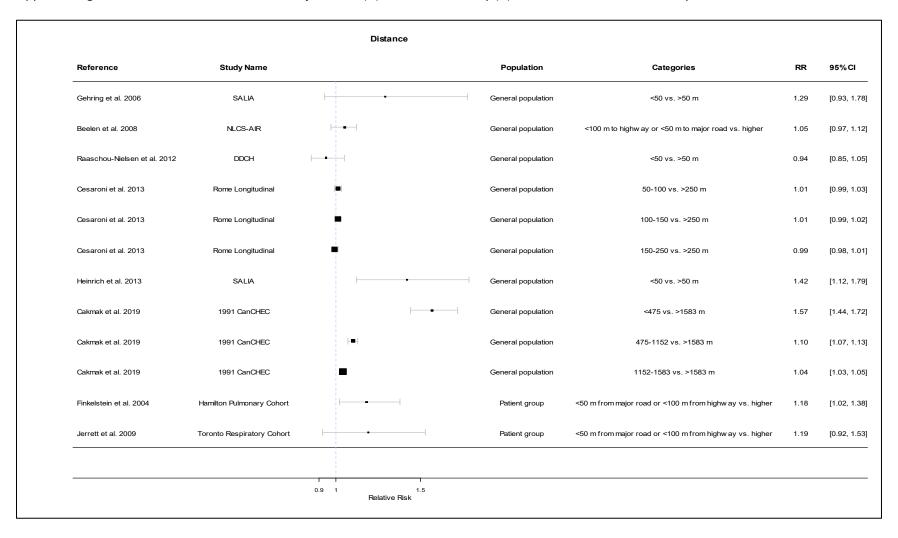
Appendix Table 8. Exposure-response assessment used for upgrading confidence for non-accidental mortality^a

Reference	Study Name	Pollutants in meta-analysis	Non-accidental mortality
Badaloni 2017	Rome Longitudinal	PM _{2.5 abs} , PM ₁₀ , PM _{2.5} , Cu, Fe	NA
Bauleo 2019	Civitavecchia Study	NO _x	NA
Beelen 2008	NLCS-AIR	NO ₂ , BC, PM _{2.5}	NA
Beelen 2014	ESCAPE	PM ₁₀ , PM _{2.5} , NO _x	+
	ESCAPE	PM _{coarse} , PM _{2.5 abs} , NO ₂	0
Beelen 2015	ESCAPE	Cu, Fe	0
Carey 2013	English National Cohort	NO ₂ , PM ₁₀ , PM _{2.5}	NA
Cesaroni 2013	Rome Longitudinal	NO ₂	+
Crouse 2015	1991 CanCHEC	NO ₂	+
Dirgawati 2019	HIMS	NO ₂ , NO _x , PM _{2.5} abs, PM _{2.5}	+
Hanigan 2019	45 and Up Study	NO ₂ , PM _{2.5}	0
Hansell 2016	ONS-Longitudinal	BS	+
	ONS-Longitudinal	PM ₁₀	NA
Hvidtfeldt 2019	DDCH	NO ₂ , BC, PM ₁₀ , PM _{2.5}	+
Krewski 2009	ACS-CPS II LA	PM _{2.5}	NA
	ACS-CPS II NYC	PM _{2.5}	NA
Nafstad 2004	Oslo men's cohort	NO _x	+
Nieuwenhuijsen 2018	Barcelona Mega Cohort	NO ₂ , PM _{2.5 abs} , PM ₁₀ , PM _{2.5}	NA
Ostro 2015	California Teachers Study	EC, PM _{2.5} , Cu, Fe	NA
Stockfelt 2015	PPS	NO _x	+
Yang 2018	Hong Kong Elderly	BC, PM _{2.5}	NA
		NO ₂	0
Yap 2012	Renfrew/Paisley	BS	NA
	Collaborative cohorts	BS	
Yorifuji 2013	Shizuoka Elderly	NO ₂	NA

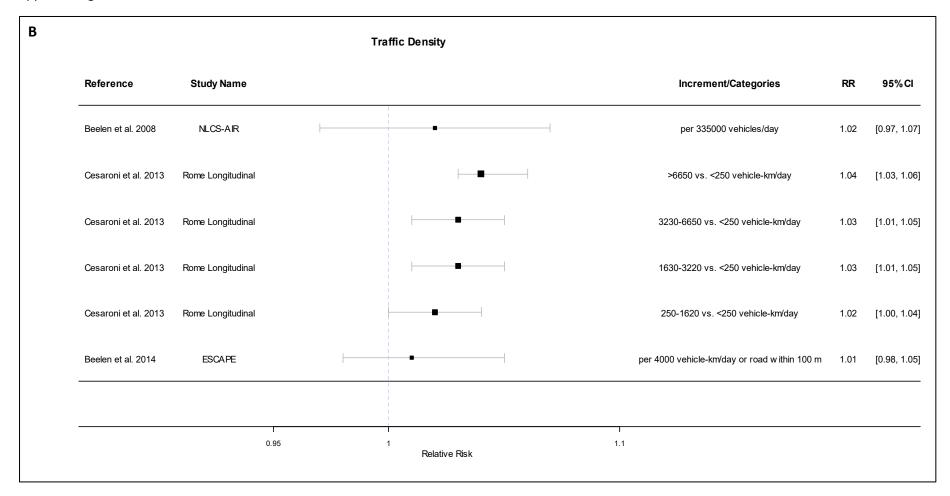
⁺ indicates evidence of a monotonic exposure—response function; 0 absence of evidence; NA: not available.

^a The Panel first assessed evidence from nonparametric spline functions, supplemented with a statistical test of deviation from linearity when available. If splines were not presented, the Panel assessed categorial exposure analyses and required a convincing trend, preferably supported by a trend test to support a judgment of a plausible exposure—response function. Finally, the Panel accepted a statement of no deviation from a linear function in the text obtained with an appropriate nonparametric procedure. To avoid upgrading null findings, the Panel only considered *no deviation from linear* as support if the linear association was at least borderline significant.

Appendix Figure 1. Association of distance to major roads (A) and traffic density (B) with non-accidental mortality.

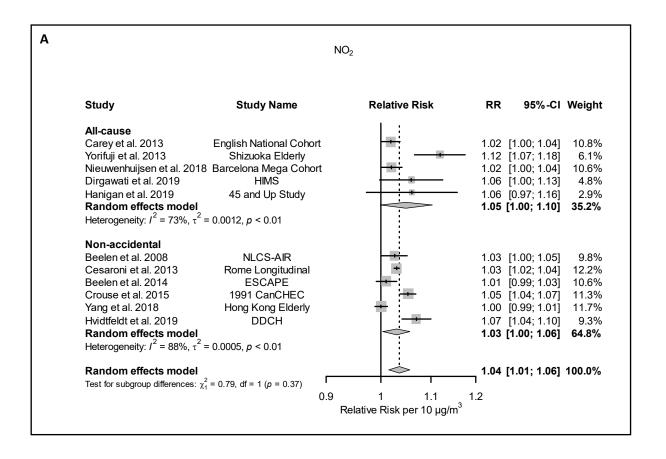


Appendix Figure 1. Continued.

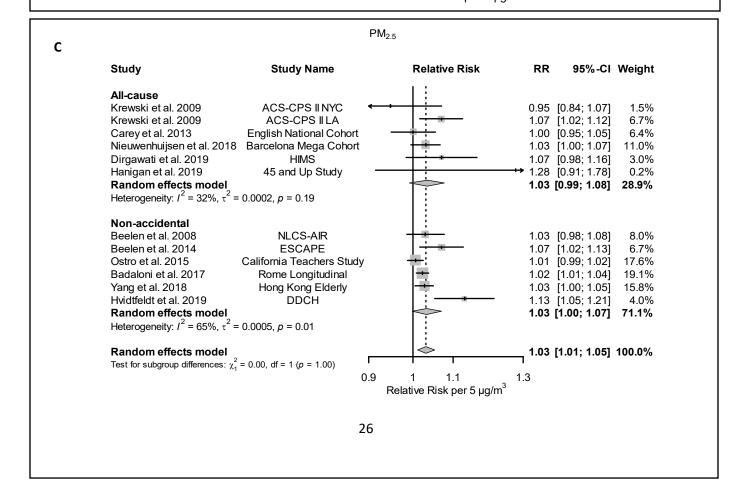


Medina-Ramón et al. 2008, Raaschou-Nielsen et al. (2012) (density) and Wilker et al. (2013) (distance) are not displayed in the plot because the estimates were log-transformed. The traffic density studies were all general population studies.

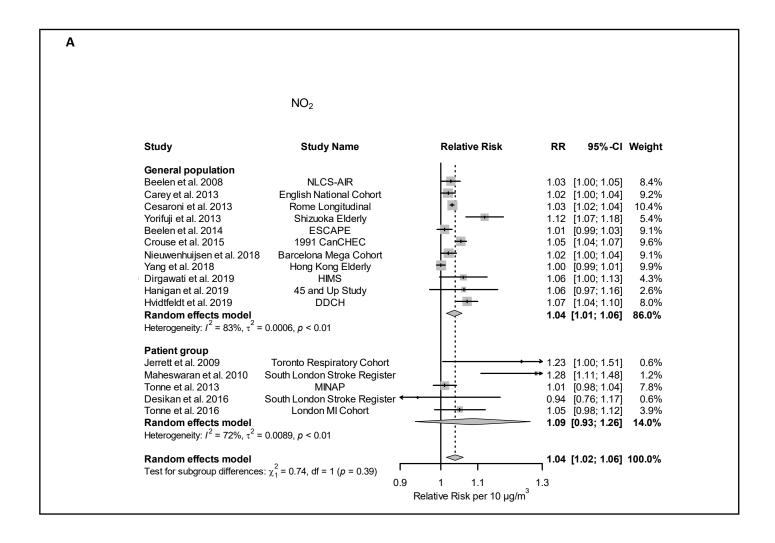
Appendix Figure 2. Association between NO_2 (A), EC (B), and $PM_{2.5}$ (C) and mortality: meta-analysis by outcome definition (all-cause versus non-accidental).



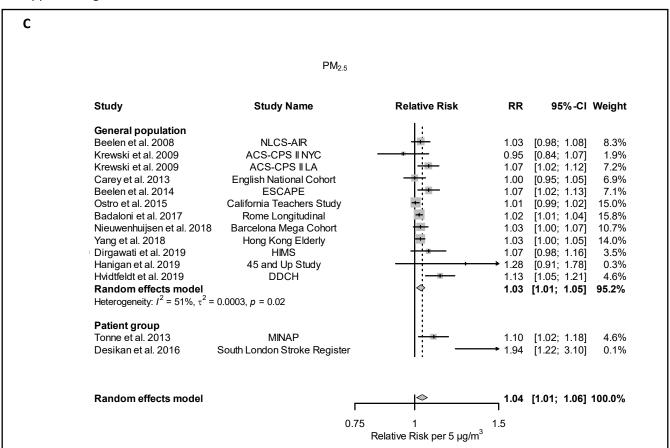
В EC Study **Study Name** Relative Risk RR 95%-CI Weight All-cause Yap et al. 2012 Renfrew/Paisley 1.07 [1.02; 1.13] 3.9% Yap et al. 2012 Collaborative cohorts 1.01 [0.96; 1.06] 4.5% Nieuwenhuijsen et al. 2018 Barcelona Mega Cohort 1.02 [1.00; 1.04] 10.8% Dirgawati et al. 2019 HIMS 1.11 [1.02; 1.20] 2.0% 1.04 [0.98; 1.11] Random effects model 21.3% Heterogeneity: $I^2 = 56\%$, $\tau^2 = 0.0008$, p = 0.08Non-accidental Beelen et al. 2008 **NLCS-AIR** 1.00 [1.00; 1.01] 16.3% 1.02 [0.97; 1.06] Beelen et al. 2014 **ESCAPE** 5.3% Ostro et al. 2015 California Teachers Study 1.00 [0.96; 1.04] 5.4% Hansell et al. 2016 ONS-Longitudinal 1.02 [1.00; 1.03] 13.9% Rome Longitudinal 1.03 [1.02; 1.04] Badaloni et al. 2017 15.3% Hong Kong Elderly Yang et al. 2018 1.00 [1.00; 1.00] 16.5% 1.07 [1.03; 1.12] Hvidtfeldt et al. 2019 DDCH 6.0% Random effects model 1.02 [1.00; 1.04] 78.7% Heterogeneity: $I^2 = 88\%$, $\tau^2 = 0.0002$, p < 0.01Random effects model 1.02 [1.00; 1.04] 100.0% Test for subgroup differences: $\chi_1^2 = 1.17$, df = 1 (p = 0.28) 0.9 1 1.1 1.3 Relative Risk per 1 µg/m³



Appendix Figure 3. Association between NO_2 (A), EC (B), and $PM_{2.5}$ (C) and non-accidental mortality: meta-analysis in general population and patient groups.



!	Study General population Beelen et al. 2008	Study Name	pollutant	Relative Risk	RR		
,	Beelen et al. 2008				1313	95% <i>-</i> CI	Weigh
,				11			
•		NLCS-AIR	BC	•	1.00	[1.00; 1.01]	16.1%
	Yap et al. 2012	Renfrew/Paisley	BS	 •	1.07	[1.02; 1.13]	4.0%
	Yap et al. 2012	Collaborative cohorts	BS		1.01	[0.96; 1.06]	4.6%
	Beelen et al. 2014	ESCAPE	PM2.5 abs	 •		[0.97; 1.06]	5.3%
		California Teachers Study		- 1		[0.96; 1.04]	5.5%
	Hansell et al. 2016	ONS-Longitudinal	BS	-		[1.00; 1.03]	13.9%
ļ	Badaloni et al. 2017	Rome Longitudinal	PM2.5 abs	 	1.03	[1.02; 1.04]	15.2%
	Nieuwenhuijsen et al. 2018	Barcelona Mega Cohort	PM2.5 abs	1		[1.00; 1.04]	10.8%
	Yang et al. 2018	Hong Kong Elderly	BC	.		[1.00; 1.00]	16.3%
	Dirgawati et al. 2019	HIMS	PM2.5 abs			[1.02; 1.20]	2.1%
-	Hvidtfeldt et al. 2019	DDCH	BC	:		[1.03; 1.12]	6.1%
	Random effects model			 	1.02	[1.00; 1.04]	99.9%
I	Heterogeneity: $I^2 = 84\%$, $\tau^2 =$	0.0002, <i>p</i> < 0.01					
ı	Patient group						
,	von Klot et al. 2009	Worcester Heart Attack	EC	: 	→ 1.79	[1.12; 2.86]	0.19
ı	Random effects model			÷	1.02	[1.00; 1.04]	100.0%
				1 1		. , .	
			0.9	1 1.1 Relative Risk per 1 μg/m	1.3		



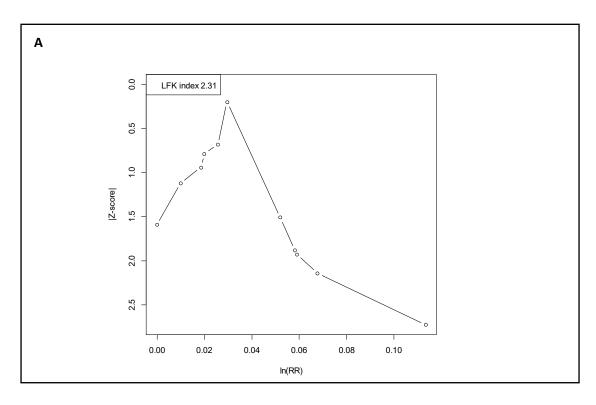
Appendix Figure 4. Association between NO_2 (A), EC (B), and $PM_{2.5}$ (C) and non-accidental mortality: meta-analysis by region.

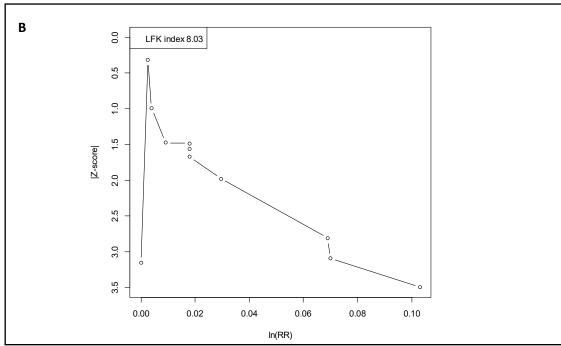
А	NO_2			
Study	Study Name	Relative Risk	RR 95%-CI	Weigh
Northern America		1 :		
Crouse et al. 2015	1991 CanCHEC	;=	1.05 [1.04; 1.07]	11.3%
Western Europe				
Beelen et al. 2008	NLCS-AIR	<u> </u>	1.03 [1.00; 1.05]	9.8%
Carey et al. 2013	English National Cohort	 	1.02 [1.00; 1.04]	10.8%
Cesaroni et al. 2013	Rome Longitudinal	+	1.03 [1.02; 1.04]	12.2%
Beelen et al. 2014	ESCAPE		1.01 [0.99; 1.03]	10.6%
Nieuwenhuijsen et al. 20	18 Barcelona Mega Cohort	 :	1.02 [1.00; 1.04]	10.6%
Hvidtfeldt et al. 2019	DDCH	i 	1.07 [1.04; 1.10]	9.3%
Random effects model			1.03 [1.01; 1.05]	63.3%
Heterogeneity: $I^2 = 61\%$, τ^2	$p^2 = 0.0002, p = 0.02$		- / -	
Asia				
Yorifuji et al. 2013	Shizuoka Elderly	1:	1.12 [1.07; 1.18]	6.1%
Yang et al. 2018	Hong Kong Elderly	= :	1.00 [0.99; 1.01]	11.7%
		1 :		
Australia/New Zealand			4.00 [4.00, 4.40]	4.00/
Dirgawati et al. 2019	HIMS		1.06 [1.00; 1.13]	4.8%
Hanigan et al. 2019	45 and Up Study	T : -	1.06 [0.97; 1.16]	2.9%
Random effects model	Г	; \$\infty	1.04 [1.01; 1.06]	100.0%
	0.9	1 1.1	1.2	
		Relative Risk per 10 µg/m ³		

Study	Study Name	Relative Risk	RR	95% -CI	Weight
Northern America Ostro et al. 2015	California Teachers Study	- :	1.00	[0.96; 1.04]	5.4%
Hvidtfeldt et al. 2019 Random effects model Heterogeneity: $I^2 = 84\%$, τ^2	NLCS-AIR Renfrew/Paisley Collaborative cohorts ESCAPE ONS-Longitudinal Rome Longitudinal 8 Barcelona Mega Cohort DDCH = 0.0002, p < 0.01	+ + + +	1.07 1.01 1.02 1.02 1.03 1.02 1.07	[1.00; 1.01] [1.02; 1.13] [0.96; 1.06] [0.97; 1.06] [1.00; 1.03] [1.02; 1.04] [1.00; 1.04] [1.03; 1.12] [1.01; 1.04]	10.8% 6.0%
Asia Yang et al. 2018	Hong Kong Elderly		1.00	[1.00; 1.00]	16.5%
Australia/New Zealand Dirgawati et al. 2019	HIMS		- 1.11	[1.02; 1.20]	2.0%
Random effects model		0.9 1 1.1 Relative Risk per 1 µg/m ³	1.02	[1.00; 1.04]	100.0%

	Р	M _{2.5}	
Study	Study Name	Relative Risk	RR 95%-Cl Weight
Northern America		1:	
Krewski et al. 2009	ACS-CPS II NYC		0.95 [0.84; 1.07] 1.5%
Krewski et al. 2009	ACS-CPS II LA	; 	1.07 [1.02; 1.12] 6.7%
Ostro et al. 2015	California Teachers Study	严	1.01 [0.99; 1.02] 17.6%
Random effects model			1.02 [0.90; 1.15] 25.9%
Heterogeneity: $I^2 = 67\%$, $\tau^2 =$	0.0014, p = 0.05		
Western Europe		li l	
Beelen et al. 2008	NLCS-AIR	 	1.03 [0.98; 1.08] 8.0%
Carey et al. 2013	English National Cohort	- † :	1.00 [0.95; 1.05] 6.4%
Beelen et al. 2014	ESCAPE	: =	1.07 [1.02; 1.13] 6.7%
Badaloni et al. 2017	Rome Longitudinal	+	1.02 [1.01; 1.04] 19.1%
Nieuwenhuijsen et al. 2018	Barcelona Mega Cohort	 •	1.03 [1.00; 1.07] 11.0%
Hvidtfeldt et al. 2019	DDCH	:	1.13 [1.05; 1.21] 4.0%
Random effects model		⊳	1.04 [1.00; 1.07] 55.1%
Heterogeneity: $I^2 = 53\%$, $\tau^2 =$	0.0003, p = 0.06		
Asia		<u> i</u>	
Yang et al. 2018	Hong Kong Elderly	ļ j	1.03 [1.00; 1.05] 15.8%
		I i	
Australia/New Zealand		[<u>[</u>	4.07.10.00.4.40
Dirgawati et al. 2019	HIMS		1.07 [0.98; 1.16] 3.0%
Hanigan et al. 2019	45 and Up Study	-1:	1.28 [0.91; 1.78] 0.2%
		ļį.	
Random effects model		→	1.03 [1.01; 1.05] 100.0%
	0.75	1 Relative Risk per 5 μg/m³	1.5

Appendix Figure 5. Doi plots and LFK indices for NO_2 (A), EC (B), and $PM_{2.5}$ (C) and non-accidental mortality.





Appendix Figure 5. Continued.

