

ORIGINAL ARTICLE

Associations of short-term exposure to traffic-related air pollution with cardiovascular and respiratory hospital admissions in London, U.K.

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What this paper adds

- Little is known about the relative contribution of the sources and constituents to traffic-related exposure health effects.
- We selected metrics, from an extensive database, indicative of traffic sources.
- Exhaust related metrics were associated with adult (15-64 years) cardiovascular and paediatric (0-14 years) respiratory hospitalizations.
- Aluminium, mineral dust tracer, was associated with adult cardiovascular admissions and respiratory hospitalizations mainly among those >15 years.
- Multi pollutant models indicate independence of associations from different sources.

ABSTRACT

Objectives. There is evidence of adverse associations between short-term exposure to traffic-related pollution and health, but little is known about the relative contribution of the various sources and particulate constituents.

Methods. For each day for 2011-12 in London, U.K. over 100 air pollutant metrics were assembled using monitors, modelling and chemical analyses. We selected *a priori* metrics indicative of traffic sources: general traffic, petrol exhaust, diesel exhaust and non-exhaust (mineral dust, brake and tyre wear). Using Poisson regression models, controlling for time varying confounders, we derived effect estimates for cardiovascular and respiratory hospital admissions at pre-specified lags and evaluated the sensitivity of estimates to multi-pollutant modelling and effect modification by season.

Results. For single day exposure, we found consistent associations between adult (15-64 years) cardiovascular and paediatric (0-14 years) respiratory admissions with elemental and black carbon (EC/BC), ranging from 0.56% to 1.65% increase per interquartile range change, and to a lesser degree with carbon monoxide (CO) and aluminium (Al). The average of past seven days EC/BC exposure was associated with elderly (65+years) cardiovascular admissions. Indicated associations were higher during the warm period of the year. Although effect estimates were sensitive to the adjustment for other pollutants they remained consistent in direction, indicating independence of associations from different sources, especially between diesel and petrol engines, as well as mineral dust.

Conclusions. Our results suggest that exhaust related pollutants are associated with increased numbers of adult cardiovascular and paediatric respiratory hospitalizations. More extensive monitoring in urban centres is required to further elucidate the associations.

INTRODUCTION

Epidemiological research has provided ample evidence for the adverse health effects of outdoor air pollution, mostly related to particulate pollution.[1] Nevertheless, there remain significant gaps in our understanding of the most harmful constituents of ambient particles and their sources.[1,2] In urban areas, traffic-related pollution, comprising primary exhaust emissions from motor vehicles, road abrasion and tyre and brake wear, is of particular concern.[1]

Whilst earlier epidemiological studies had identified associations between residence proximity to busy roads with outcomes such as cardiovascular and respiratory mortality,[3] cardiovascular disease (CVD),[4] lung function,[5] the large scale European Study of Cohorts for Air Pollution Effects (ESCAPE) using near traffic exposure metrics such as PM absorbance, nitrogen oxides (NO_x) or traffic load and intensity failed to confirm associations with mortality,[6] incidence of lung cancer,[7] cerebrovascular[8] or acute coronary events.[9] Instead, this multi-cohort study provided evidence that traffic exposure metrics were associated with adverse paediatric respiratory outcomes,[10] and elevated blood pressure or prevalent hypertension among adults.[11] Epidemiological time series studies of short term exposure and health effects have also reported mixed results for a range of health endpoints with individual pollutants, such as particles with aerodynamic diameter less than 2.5 µm (PM_{2.5}), nitrogen dioxide (NO₂), carbon monoxide (CO) or black carbon (BC).[11-17] A previous study in London, U.K.[18] suggested that certain particle components might be more important to specific diseases, pointing to particle number concentrations for CVD and secondary pollutants for respiratory outcomes. The biological mechanism for the traffic-related associations remains poorly understood, although some toxicological studies have suggested pulmonary and vascular inflammation as the relevant mechanism.[1,2]

Previous investigation of the relative contribution of pollutants and sources using daily time-series analysis methods has been limited by dependence on data from routine monitors.[1] The Clearflo project[19] characterised, in detail, the air pollution mixture in London between 2011-2012 and provided the opportunity to conduct daily time-series analyses focusing on specific sources, utilising data on the chemical composition of particles, estimation of the urban increment, as well as routine and study specific pollutant measurements. We selected *a priori* those metrics that best represented general traffic sources, diesel and petrol combustion, and non-exhaust sources (brake, tyre and road re-suspension) for inclusion in a time series analysis of respiratory and CVD daily emergency hospital admissions.

DATA AND METHODS

Data

Daily counts of emergency hospital admissions in London, U.K. between 2011-12 were constructed from individual records of hospital admission obtained from the Hospital Episode Statistics (HES) system. Outpatient visits, elective admissions and visits to emergency department were not included. Based upon the primary discharge diagnosis, daily numbers of admissions for CVD (International Classification of Diseases, 10th revision – ICD10: I00-I99) for those aged 15-64 (adult) and 65+ years (elderly), and respiratory diseases (ICD10: J00-J99) for those aged 0-14 years (paediatric), adult and the elderly were calculated.

Using data collected from the Clearflo project,[19] supplemented by local measurements made at the North Kensington urban background site, we assembled a database of metrics for 2011-2012, that

included daily concentrations of particle mass (for particles with aerodynamic diameter less than 10 μm (PM_{10}), or $\text{PM}_{2.5}$), as well as particle composition (carbon, anions and metals) and gases (NO_2 , NO_x , CO, sulphur dioxide (SO_2) and ozone (O_3)). All concentrations were based on 24h averages except for CO and O_3 for which the maximum 8h moving average was computed. Daily concentrations of NO_x , CO, BC and EC attributable to London sources rather than air mass transport were estimated by calculating the urban increment between North Kensington and two monitoring sites in the rural area around London dependent upon wind direction each day.[20]

We then adopted a hypothesis driven approach to the analyses. Based upon a review of the literature on source identification, the London atmospheric emissions inventory[21] and analysis of temporal trends and correlations we selected, *a priori*, pollutants to represent specific traffic sources (see Supplemental Material Annexes 1-3 for detailed description and justification of selected pollutants, including correlation coefficients). In brief: (1) NO_x was selected as a general traffic indicator, as 47% of it is emitted by road transport;[21] (2) CO was selected as a proxy for emissions from petrol vehicles in London, as the contribution from petrol cars ranges from 0.07% to 0.9%, as compared to 0.01% to 0.07% for diesel vehicles;[22] (3) Elemental carbon (EC) in PM_{10} and black carbon (BC) in $\text{PM}_{2.5}$ were chosen as indicators of diesel exhaust as studies of real-world vehicle emissions in London have demonstrated that diesel vehicles are overwhelmingly the largest emitters of EC and BC;[23] (4) copper (Cu) in PM_{10} was selected as the indicator of brake-generated particles, as it is the most abundant element in brake linings and is found in high abundance in brake dust;[24] (5) zinc (Zn) to reflect tyre-generated particles;[24] and (6) aluminium (Al) as a marker of dust re-suspension, as it occurs in sufficient quantities and is not identified in other sources.[25] Regulated pollutants (PM_{10} , $\text{PM}_{2.5}$, NO_2 , SO_2 and O_3) were also selected for comparability with previous findings and mutual

control in multi-pollutants' models. There were few missing values in the pollution time-series (ranging from 0% for particle mass concentrations and CO to 19% for EC urban increment).

Time series of daily temperature (°C, mean) and relative humidity (%) were obtained from a meteorological tower located close to the North Kensington monitoring site.

Methods

We investigated the associations between short-term exposure to traffic-related pollutants and daily hospital admissions using Poisson regression models allowing for overdispersion. The model was of the form:

$$\log E [Y_t] = \beta_0 + b * Pol_t + s(time_t) + \sum_i s(X_{it})$$

where $E[Y_t]$ is the expected value of the Poisson distributed variable Y_t indicating the daily outcome count on day t with $Var(Y_t) = \phi E[Y_t]$, ϕ being the over-dispersion parameter, $time_t$ is a continuous variable indicating the time (day) of event (from 1 to 731), Pol_t is the pollutant concentration on day t , X_{it} is the value of confounder X_i on day t , and s denotes smoothing functions. We used penalized regression splines[26] as smoothing functions s to capture the association between time-varying covariates, calendar time and health outcome. Degrees of freedom (df) for long term trends were based on the minimization of the absolute value of the sum of the partial autocorrelations (PACF) of the residuals from lags 1 to 30, imposing a minimum of 3 df per year. We also included dummy variables for the day of the week and public holidays. For the analysis of respiratory admissions among ages 0-14 and 15-64 years we included an extra dummy variable denoting the month of August, as the decrease in the respiratory admissions at this period could not be sufficiently captured by the smooth term of seasonality. We controlled for mean daily temperature and relative humidity to address any

potential confounding effects of weather. For temperature control we applied a natural spline with 3 df for same day's exposure (lag 0) to capture the effect of high temperatures on health, while to capture the health effects of lower temperatures we used the corresponding function on the average of the six previous days exposure (lags 1-6), as these terms minimized the Akaike's Information Criterion. For relative humidity adjustment, we included a linear term for the average of the same and the two previous days, sufficient to capture any residual weather confounding. When we investigated the associations with EC/BC and metal components of particles, we also controlled for particle mass (PM₁₀ for EC and metals and PM_{2.5} for BC), as a way to distinguish the effect of the particular constituent from the rest.[27]

We decided *a-priori* which lags of the pollutants to be included in the models: previous day's exposure for CVD admissions (lag1) and previous two days' exposure for respiratory admissions (lag2), based on prior indications of longer lags for respiratory outcomes.[18] To investigate any prolonged effects, we additionally applied unconstrained distributed lag (DL) models for the previous week's exposure (lags 0-6).

We applied multi pollutant models after considering the correlations between pollutant pairs (see Supplemental Material Annex 3). We included pollutants in a model in cases when the correlation was below 0.7. Specifically we applied two pollutant models to test the robustness of the associations with gases and three pollutant models for EC/BC and metal components of particles. For gases, the second pollutant entered in the model was selected in order to test the hypothesis of independent effects between traffic or long range transport related metrics (NO_x or CO controlling for PM_{2.5} or EC, SO₂ and O₃). For EC/BC and metals, for which already the corresponding particle mass was controlled in the model, we additionally adjusted for NO_x and CO. In order to minimize the correlation between the three metrics, instead of adding the third metric in the model, we initially regressed the gaseous

pollutant on particle mass (PM₁₀ or PM_{2.5}) and consequently entered the model residuals in the model to adjust for any remaining effect not attributed to particles.[27]

We investigated the associations by season defined as warm (April-September) and cool (October-March) period to test the hypothesis of effect modification due to differential sources and exposure misclassification between periods. For these analyses we controlled for seasonality and long-term trends using indicator variables per month per year of the study, while the rest of the confounding control was the same as in the annual model.

All models were fit in R v.3.0.3 (R development Core Team (2011), ISBN 3-900051-07-0, URL <http://www.R-project.org>) using the package *mgcv* (v.1.7-28). Results in tables and plots are presented as percent change associated with an interquartile increase (IQR) in the pollutant's concentration.

RESULTS

Table 1 presents descriptive statistics for the daily number of hospital admissions, daily concentrations for the pollutants and meteorological parameters. The greater London area had a population of 9,787,426 inhabitants (2011 Census). The mean number of hospital admissions per day varied from 104 for CVD in the elderly to 46 for paediatric respiratory diseases. Mean particulate matter concentrations were 18.4µg/m³ for PM₁₀ and 12.2 µg/m³ for PM_{2.5}, while mean concentrations of gaseous pollutants were 55.3 µg/m³ for NO_x, 1.8 µg/m³ for SO₂ and 0.3mg/m³ for CO. The urban increment of NO_x, CO, EC and BC accounted for most of the measured concentration, showing them to be dominated by urban sources, with the exception of CO (mean concentration 0.3 mg/m³, with estimated urban increment of 0.1 mg/m³). Higher concentrations of traffic-related pollutants were

recorded during the cool period (see Supplemental Material Annex 3), however the roadside enrichment factors were lower, when compared with the warm period. Cool period enrichment factors for NO_x, BC, EC and CO were 3.5, 4.5, 4.2 and 1.1 respectively, increasing to 6.8, 7.2, 7.3 and 1.8 during the warm months, implying that roadside sources were more dominant in the warm period, even though total pollutant concentrations were lower. Overall period correlations among pollutants ranged from 0.2 (correlations with O₃) to greater than 0.9 (see Supplemental Material Annex 3). Specifically, correlations of CO were: 0.83 with NO_x, 0.77 with BC and 0.62 with Cu. Correlations using the urban increment of the pollutants were substantially smaller than those with the total measured concentration; e.g. the correlation between NO_x and CO was 0.83, but was reduced to 0.41 when only the urban increment was considered.

Table 1. Descriptive characteristics of hospital admissions counts, traffic-related pollutants and meteorological variables in London, U.K. for 2011–12.

	Number of days	Mean	Median	IQR ^a (75 th -25 th percentile)	90 th percentile
Hospital Admissions					
Cardiovascular					
<i>15-64 years</i>	731	56	57	25	71
<i>65+ years</i>	731	102	104	37	124
Respiratory					
<i>0-14 years</i>	731	46	45	23	72
<i>15-64 years</i>	731	64	63	16	81

<i>65+ years</i>	731	96	91	28	125
Pollutants ($\mu\text{g}/\text{m}^3$; CO in mg/m^3)					
General traffic indicator					
<i>NO_x</i>	706	55.3	41.2	41.3	106.5
<i>NO_x Urban increment</i>	703	42.5	30.8	33.1	84.4
Petrol vehicle exhaust					
<i>CO</i>	729	0.3	0.3	0.2	0.5
<i>CO Urban Increment</i>	724	0.10	0.08	0.09	0.21
Diesel vehicle exhaust					
<i>EC (in PM₁₀)</i>	682	1.0	0.8	0.8	1.9
<i>EC Urban (in PM₁₀)</i>	590	0.8	0.6	0.5	1.4
<i>BC (in PM_{2.5})</i>	702	1.5	1.2	1	2.8
<i>BC Urban (in PM_{2.5})</i>	629	0.9	0.7	0.6	1.8
Vehicle Non-exhaust					
<i>Cu (in PM₁₀)</i>	677	0.0093	0.0072	0.0075	0.0176
<i>Zn (in PM₁₀)</i>	677	0.012	0.0087	0.0091	0.0246
<i>Al (in PM₁₀)</i>	677	0.076	0.0555	0.0605	0.1528
Regulated Pollutants ($\mu\text{g}/\text{m}^3$)					
<i>PM₁₀</i>	729	18.4	15.0	10	32.5
<i>PM_{2.5}</i>	730	12.2	9.0	8	25.0
<i>NO₂</i>	706	36.3	33.3	23.7	58.1
<i>SO₂</i>	717	1.8	1.8	2.2	3.6
<i>O₃</i>	716	55.4	54.7	30.3	85.9

Meteorological Parameters

<i>Mean Temperature (°C)</i>	731	11.70	11.70	7.5	18.10
<i>Relative humidity (%)</i>	731	76.43	78.00	14.6	88.50

^a IQR: Inter-Quartile Range

Table 2 presents the percent change in hospital admissions for an IQR increase in the concentrations of the traffic-related pollutants following single day exposure (lag1 for CVD and lag2 for respiratory diagnoses). Associations with regulated pollutants are presented in Supplemental Material Annex 4 as these were not the focus of the present analysis. Table 3 presents the percent change following weekly exposure (lags 0-6).

Table 2. Percent change (and 95% confidence intervals (CIs)) in cardiovascular and respiratory hospital admissions associated with an interquartile range increase in traffic-related pollutants after acute exposure (lag 1 for cardiovascular and lag 2 for respiratory diagnoses) in London, U.K. for 2011–12. EC/BC and metals are adjusted for PM mass.

Indicator/ Pollutants	CVD Admissions % (95% CI)		Respiratory Admissions % (95% CI)		
	15-64 years	65+ years	0-14 years	15-64 years	65+ years
General traffic					
<i>NOx</i>	0.86 (-0.28, 2.02)	-0.32 (-1.19, 0.56)	1.06 (-0.43, 2.57)	-0.81 (-1.92, 0.31)	-1.76 (-2.77, -0.74)
<i>NOx – Urban</i>	0.92 (-0.15, 2.00)	-0.15 (-0.97, 0.67)	1.07 (-0.31, 2.46)	-0.60 (-1.64, 0.46)	-1.51 (-2.45, -0.55)
Petrol vehicle exhaust					
<i>CO</i>	1.59 (0.12, 3.07)	-0.60 (-1.71, 0.52)	1.05 (-0.96, 3.10)	-1.11 (-2.57, 0.36)	-2.10 (-3.43, -0.75)
<i>CO –Urban</i>	0.95 (-0.06, 1.98)	-0.16 (-0.93, 0.62)	0.97 (-0.40, 2.36)	0.21 (-0.80, 1.24)	-0.59 (-1.52, 0.34)
Diesel vehicle exhaust					
<i>EC</i>	1.63 (0.15, 3.13)	0.18 (-0.94, 1.32)	0.72 (-1.22, 2.70)	-0.19 (-1.63, 1.27)	-0.88 (-2.19, 0.45)
<i>EC - Urban</i>	1.28 (0.17, 2.40)	0.14 (-0.72, 1.00)	1.27 (-0.21, 2.78)	-0.03 (-1.12, 1.08)	-0.05 (-1.04, 0.96)
<i>BC</i>	1.65 (0.11, 3.21)	0.56 (-0.61, 1.74)	0.86 (-1.13, 2.88)	-0.20 (-1.71, 1.33)	-1.09 (-2.47, 0.31)

<i>BC –Urban</i>	0.74 (-0.47, 1.97)	-0.04 (-0.97, 0.89)	1.08 (-0.50, 2.68)	0.44 (-0.78, 1.66)	-0.01 (-1.10, 1.09)
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Vehicle Non-exhaust

<i>Cu</i>	1.39 (-0.03, 2.83)	0.06 (-1.02, 1.16)	0.08 (-1.81, 2.01)	-1.18 (-2.60, 0.26)	-1.60 (-2.89, -0.28)
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<i>Zn</i>	0.08 (-1.25, 1.42)	0.16 (-0.85, 1.18)	-0.92 (-2.72, 1.47)	-0.38 (-1.73, 1.00)	-0.73 (-1.96, 0.52)
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<i>Al</i>	0.43 (-1.18, 2.07)	-1.14 (-2.35, 0.09)	0.19 (-2.22, 2.66)	0.82 (-0.84, 2.50)	1.38 (-0.15, 2.94)
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Table 3. Percent change (and 95% confidence intervals (CIs)) in cardiovascular and respiratory hospital admissions associated with an interquartile range increase in traffic-related pollutants after weekly exposure (lags 0-6) in London, U.K. for 2011–12. EC/BC and metals are adjusted for PM mass.

Indicator/Pollutants	CVD Admissions % (95% CI)		Respiratory Admissions % (95% CI)		
	15-64 years	65+ years	0-14 years	15-64 years	65+ years
General traffic					
<i>NOx</i>	-0.92 (-2.98, 1.18)	0.20 (-1.38, 1.80)	4.01 (0.76, 7.37)	-1.67 (-3.70, 0.39)	-5.13 (-6.98, -3.24)
<i>NOx – Urban</i>	-0.37 (-2.43, 1.73)	0.45 (-1.12, 2.04)	3.86 (0.67, 7.16)	-0.95 (-2.97, 1.11)	-4.77 (-6.61, -2.90)
Petrol vehicle exhaust					
<i>CO</i>	1.03 (-1.85, 3.99)	-1.18 (-3.31, 1.01)	4.94 (0.11, 10.00)	-2.51 (-5.29, 0.36)	-8.81 (-11.28, -6.28)
<i>CO – Urban</i>	2.52 (0.17, 4.92)	-1.08 (-2.82, 0.69)	4.02 (0.24, 7.93)	-0.31 (-2.64, 2.07)	-3.68 (-5.87, -1.43)
Diesel vehicle exhaust					
<i>EC</i>	1.39 (-1.59, 4.45)	2.36 (0.05, 4.73)	1.64 (-2.85, 6.35)	-1.62 (-4.54, 1.39)	-2.97 (-5.69, -0.18)
<i>EC - Urban</i>	1.46 (-0.93, 3.91)	1.65 (-0.18, 3.52)	2.59 (-1.45, 6.79)	0.56 (-1.96, 3.14)	-0.02 (-2.38, 2.40)
<i>BC</i>	0.13 (-2.84, 3.19)	1.49 (-0.82, 3.86)	4.01 (-0.70, 8.94)	-0.87 (-3.84, 2.21)	-3.11 (-5.91, -0.23)

<i>BC –Urban</i>	0.49 (-3.00, 4.11)	2.13 (-0.57,4.91)	1.34 (-3.76, 6.70)	0.64 (-2.88, 4.29)	-0.40 (-3.61, 2.91)
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Vehicle Non-exhaust

<i>Cu</i>	2.23 (-0.39, 4.91)	1.11 (-0.88, 3.14)	-5.60 (-9.28, -1.77)	-1.98 (-4.58, 0.70)	-2.13 (-4.57, 0.36)
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<i>Zn</i>	0.91 (-1.96, 3.87)	0.12 (-2.08, 2.37)	-4.26 (-8.24, -0.11)	-3.17 (-6.00, -0.25)	-2.53 (-5.15, 0.16)
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<i>Al</i>	0.75 (-2.01, 3.59)	0.32 (-1.78, 2.47)	-6.43 (-11.08, -1.55)	2.46 (-0.50, 5.50)	4.49 (1.62, 7.473)
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General traffic indicator

Both the total measured concentration at North Kensington and the urban increment of NO_x displayed positive associations (Table 2) with CVD adult admissions and paediatric respiratory admissions. Moreover, when considering longer lags of exposure (lags 0-6, Table 3) NO_x presented an adverse association with CVD admissions in the elderly, 0.20% increase (95% Confidence Interval (CI): -1.38%, 1.80%) for an IQR increase in total measured NO_x and 0.45% increase (95% CI: -1.12%, 2.04%) for its urban increment. Associations between measured concentrations of NO_x, and its urban increment, and paediatric respiratory hospital admissions were also observed, 4.01% (95% CI: 0.76%, 7.37%) and 3.86% (95% CI: 0.67%, 7.16%) respectively. Negative associations, for both the total and the urban part concentration, irrespective of the lag structure, were observed for respiratory admissions in the elderly population.

Petrol vehicle exhaust indicator

CO, total concentration and urban increment, was associated with CVD adult admissions, more strongly following lag 1 exposure for CO (1.59% increase (95% CI: 0.12%, 3.07%), Table 2) and lags 0-6 exposure for its urban increment (2.52% increase (95% CI: 0.17%, 4.92%), Table 3). Positive associations were also estimated for paediatric respiratory admissions that became higher for longer exposures, while the opposite patterns were observed for respiratory admissions in the elderly.

Diesel vehicle exhaust indicators

Adverse associations with all CVD outcomes were estimated for EC following acute exposure (Table 2) among adults for the total measured mass (1.63% increase, 95% CI: 0.15%, 3.13%) and for the urban increment (1.28%, 95% CI: 0.17%, 2.40%). Similar adverse associations were also observed

after prolonged exposure (lags 0-6, Table 3) among the elderly for total measured mass (2.36% increase, 95%CI: 0.05%, 4.73%) and the urban increment (1.65%, 95% CI: -0.18%, 3.52%). Positive associations were also estimated for paediatric respiratory hospital admissions (Tables 2 and 3); while adverse associations among the adult and elderly age groups were estimated only for weekly exposures to the urban increment (Table 3).

BC, both total and urban increment, also displayed positive associations with all CVD outcomes (Tables 2 and 3), except with CVD admissions in the elderly group for lag 1, for which no association was observed. Consistent positive associations were also noted for all respiratory outcomes following either acute (Table 2) or weekly (Table 3) exposure to the urban BC contribution, while exposure to the total measured BC was associated with paediatric respiratory admissions.

Vehicle Non-exhaust (mineral dust, brake and tyre wear) indicators

We found positive associations between Cu, Zn and Al in PM₁₀ and CVD hospital admissions in adults (Tables 2 and 3). Al was positively associated with respiratory hospital admissions across all age groups and exposure periods studied (Tables 2 and 3) with the single exception for lags 0-6 exposure and paediatric respiratory admissions. Conversely, we observed negative associations between Cu and respiratory hospitalizations irrespective of age and lag.

In summary, all traffic-related pollutants displayed adverse associations with CVD admissions in adults after a single day exposure, while positive effect estimates for EC and BC were observed for CVD admissions among the elderly. Considering weekly exposures, effect estimates for CVD admissions were generally lower for adults compared to the elderly. We observed positive associations with all pollutants except Zn on paediatric respiratory admissions. The highest effect estimates were for the association with the EC (1.27% increase) and BC (1.08%) urban increment concentrations. Longer

exposure resulted in higher estimates for paediatric respiratory admissions (except for metals). Only the urban increments of CO and BC, and AI displayed an association with respiratory admissions among adults, with only AI showing an association in the elderly (1.38% increase).

In general, effect estimates of traffic pollutants on hospital admissions displayed a consistent decreasing trend with increasing age, while associations were higher with CVD as compared with respiratory emergency admissions. For example an IQR increase in BC was associated 1.65% increase in CVD admissions in adults compared with a 0.56 % increase in the corresponding age group for respiratory admissions.

Effect estimates for regulated pollutants (PM₁₀, PM_{2.5}, NO₂, SO₂) also supported findings of effects only within the younger age group, namely for adult CVD and paediatric respiratory admissions (Supplemental Material Annex 4), while only exposure to O₃ was related to effects among the older age groups (CVD 65+years and respiratory >15 years).

The effect estimates derived from single pollutant models were robust to adjustment of other pollutants (Supplemental Material Annex 5). EC and AI presented the most consistent associations, while there was indication of confounding between CO and EC/BC that lowered the estimates; these nevertheless remained positive.

Figure 1 presents the percent change in hospital admissions by warm and cool period of the year for an IQR following single day exposure in traffic-related pollutants. Effect estimates (and 95% CIs) are presented in the Supplemental Material Annex 6. The seasonal analysis revealed higher effect estimates during the warmer period of the year, except for CVD admissions in the elderly, although in general there was no difference in the estimates between periods. The associations of EC with CVD admissions for adult and paediatric respiratory admissions were higher during the warm period, as was also the case for AI on the latter outcome.

DISCUSSION

We investigated associations between short-term exposure to an a-priori selection of traffic-related pollutants and CVD and respiratory hospital admissions in London, U.K. We found consistent positive associations between EC, BC and AI in PM₁₀ after a single day exposure and CVD outcomes in adults and with paediatric respiratory admissions. Furthermore, seven days averages for these pollutants were associated with CVD admissions in the elderly. These particular associations were larger in the warm period of the year compared to the cooler period although the differences did not achieve statistical significance.

The main strengths of our study are the range and the quality of the pollution metrics assembled from routine and augmented monitoring at a central urban background site, plus the capacity to address the urban increment for metrics where parallel monitoring was performed at rural locations. Assembling a database with such an extensive range of constituents is a major, and possibly unique, aspect of our study - an example of translational research linking laboratory with epidemiology. Moreover, London, due to its large size, provides the adequate number of mean daily counts necessary for the variability of the time-series. We adopted a hypothesis-driven approach to select a limited number of the available

metrics to reflect potential traffic sources to the urban air shed. This selection was informed by a detailed review of the literature, data completeness and the observed correlation structure between the available metrics. This *a-priori* selection of traffic indicators, as well as predefined lags for CVD and respiratory admissions limited the number of uninformed comparisons performed and we believe strengthens the robustness of our findings.

Although a limitation of the present study is the relatively small sample size (two years) for a time-series design, this was compensated for by the completeness of the pollutant measurements over the study period and the large baseline population. An inherent drawback of time-series studies is the misclassification induced due to the use of fixed monitors to estimate the population's exposure; in this case a single fixed site in North Kensington, in inner London. Nevertheless, previous analyses have identified positive associations using this urban background site.[23] We tested the sensitivity of the effect estimates of the regulated pollutants (PM₁₀, PM_{2.5}, NO₂, CO and O₃) obtained from the North Kensington site to the ones from the average of the daily measurements from available fixed monitoring stations scattered around London and the results were comparable, e.g. an IQR increase in NO₂ from North Kensington was associated with 1.00% (95% CI: -0.87%, 2.91%) change in CVD admissions 15-64 years, while the average of all urban and suburban background monitors was associated with a 0.77% (95% CI: (-1.01%, 2.58%)) change. Nevertheless, due to possible differences in the spatial variability of the traffic derived pollutants measured in North Kensington, there remains a chance of residual confounding, though these are not likely to rule out causation, especially in pollutants that display consistent associations.[23] Although multiple comparisons for detection of associations may have resulted in inflation of type I error, we chose not to correct for this but instead identified associations with pollutants that were consistent across different outcomes. Finally, the chosen metrics did not entirely represent traffic sources as they partly originate from other sources. The

degrees of specificity for traffic sources can be gauged from the roadside enrichment factors (Supplemental Material Annex 1) which varied from 1.3 for Al and Zn and 1.4 for CO, indicating a relatively low contribution from traffic, to 4.6 for NO_x, 4.7 for Cu and 5.6 for BC, where traffic sources made a greater contribution. The contribution from traffic also varies by season; for instance it was lower for NO_x in winter when space heating also contributed. Differential pollutant dispersion between seasons also has an effect in addition to source changes.

The urban increments of EC and BC provided lower effect estimates for CVD admissions and higher estimates for respiratory admissions compared to the corresponding total concentrations. There were indications of associations with both pollutants among people below 65 years, while more prolonged exposure was associated with CVD outcomes among the elderly (65+years). In general, the BC urban increment revealed a more consistent pattern. BC and EC were selected a-priori as diesel exhaust markers (see Supplemental Material Annex 1) and the results from the multi pollutants' analysis indicated that although there was some confounding when adjusting for CO, as a petrol exhaust marker, the adverse associations generally remained, pointing towards independent effects of different sources. Recent reviews[21,28] on BC health effects concluded that, although there is sufficient evidence on short-term exposure and effects on cardiopulmonary admissions, the toxicological evidence suggested that BC may not be a the major directly toxic component of fine PM. BC may operate as a universal carrier of a wide variety of chemicals of varying toxicity to the lungs, which may then induce adverse effects within and beyond the respiratory system, the body's major defence cells and possibly the systemic blood circulation. Alternatively, BC may act as a surrogate of true causal pollutants correlated through a common source. Recent experimental studies have demonstrated systemic effects of BC on arterial blood pressure responses[29] and of diesel exhaust itself on hemoconcentration and thrombocytosis -potentially important determinants of acute CVD events.[30]

CO, selected as a petrol indicator, was associated with an increased risk of CVD admissions among adults. The urban increment in CO also provided consistent associations with adult respiratory admissions. There was some indication of confounding with EC and BC (as diesel markers) but the association was still apparent. A meta-analysis of single or multi-cities results[31] reported evidence for adverse effects of short term CO exposures on hospitalizations due to respiratory or diagnoses-specific CVD admissions, and Bell et al.[32] also found evidence of an association with risk of CVD hospitalizations in 126 U.S. counties. Notably, although in both studies concentrations of CO were well below the E.U. air quality standards (<http://ec.europa.eu/environment/air/quality/standards.htm>) the associated health effect estimates were high compared to the other pollutants.

We found little evidence of effects of NO_x as a general traffic marker, although there were positive associations with CVD adult admissions that persisted in two pollutant models. The increase in CVD is compatible with the higher systolic blood pressure reported by Kubesch et al.[29] following short-term exposure to traffic-related air pollution. We also observed positive but not statistically significant associations with paediatric respiratory admissions (1.06% increase per IQR). Previous panel studies on asthmatic children have reported adverse effects of exposure to NO_x,[1] while Iskandar et al. [33] using a time series design also reported increases in hospital admissions for asthma among 0-18 years. On the other hand, consistent protective associations were found with respiratory admissions among the elderly. Protective associations are not supported by plausible biological mechanisms, but it is possible that elderly subjects with respiratory conditions may avoid outdoor exposure as a result of public health warning messages (air pollution forecasts are incorporated into weather forecasts in the UK) or use prophylactic medications and hence modify the associations observed in our study. Most previous times-series studies have focused on NO₂. However, NO_x may also reflect NO₂ effects due to their high correlation ($r=0.90$), further supported by their similar effect estimates. The WHO review[1]

concluded that there is consistent epidemiological evidence and some mechanistic support for causality of some NO₂ direct effects. In our analysis adjustment for particles, sulphates or gaseous regulated pollutants increased our effect estimates supporting the plausibility of the reported associations.

Cu, Zn and Al in PM₁₀ were selected as markers of non-exhaust traffic contributions to PM₁₀. Al, selected as a mineral dust tracer, demonstrated adverse associations with CVD admissions in adults (0.43% increase per IQR) and with respiratory admissions in adults (0.82% per IQR increment) and elderly (1.38% per IQR increment). Bell et al.[34] also reported effects of Al on respiratory admission among the elderly, which is the association with the highest effect estimate also in our analysis. There was also some evidence of an association between Cu as tracer of brake generated particles and CVD admissions. Basagana et al.[35] using data from five Southern European cities reported adverse associations of Cu on CVD morbidity, that were higher than the ones found in London (1.94% increase per IQR), but did not persist after adjustment for PM mass as in our data. There is toxicological evidence for the biological mechanism of effects, as Cu and Zn have both been linked to a decrease in spontaneous beat rate, vasoconstriction and vasodilatation.[36] However, we did not find convincing evidence for an association between Zn and CVD or respiratory admissions, although this metal has been studied and associations have been previously reported.[1,35]

The higher effect estimates of all traffic-related pollutants observed in the younger age groups, between 15-64 years for CVD and mainly in children 0-14 years for respiratory admissions, are of particular interest. The underlying mechanisms for the observed patterns may be attributed to age-specific different diagnoses, but also to moderation of pre-existing disease (especially CVD) in the elderly. This hypothesis is also supported by the indication of associations with CVD outcomes following longer periods of exposure. Few epidemiological studies on the effects of short-term exposure to air pollution have reported age modification patterns in hospitalizations. Using European data from the 1990s, Le

Tertre et al.[37] found larger effects of particles on CVD admissions among the elderly, but since 2000 the increased use of statins and other medications for CVD diseases could potentially have modified this risk. Higher respiratory effects of particles in younger ages have been previously reported,[38] though previous reports of particle related effects on the elderly do not chimes with our current findings.[38,39] Finally, it is also possible that as a response to increased public awareness during the last decades on the health effects of air pollution and inclusion of air pollution levels and forecasts into the daily weather forecasts, sensitive sub-groups such as the elderly with pre-existing conditions may have modified their time activity patterns resulting in modification of effects.

We found higher effect estimates between the selected traffic-related pollutants (except for CO) and adult (15-64 years) CVD admissions and respiratory admissions among those below 65 years of age during the warmer period of the year compared to the cooler period. European epidemiological studies on the health effects of short-term exposure to air pollution have reported higher associations during the warmer period of the year, [14, 39] often attributed to better exposure characterization of the population. Nevertheless US studies [40] have reported higher numbers of hospitalizations for the elderly during the cool period of the year suggesting that seasonal patterns may differ across age groups that potentially follow different activity patterns. Moreover, toxicity of particles originating from different sources may vary between seasons and locations. In the current analysis the roadside enrichment factors suggest that traffic sources are more dominant in summer months due to seasonal variation in sources and dispersion.

The results indicate consistent associations predominately with EC/BC but also to a lesser degree with CO and PM₁₀ Al content, with CVD and respiratory admissions among younger age groups. Although the specific effect estimates are variable to the adjustment of other pollutants they remain largely consistent in direction, indicating the independence of effects from different traffic sources especially

from diesel and petrol engines, as well as resuspended mineral dust. Supporting this argument, the cumulative effect of traffic related pollution is not appropriately captured from a general indicator such as NO_x, possibly due to differential patterns of pollutant correlations or associated effects. This conclusion is crucial for planning public health policies aiming at the reduction of air pollution effects.

Our findings point towards short-term exposure to exhaust rather than non-exhaust related pollutants as the ones mostly associated with adverse effects on morbidity, previously attributed to traffic-related pollutants. As diesel powered engines are the main urban source of EC and BC, which presented the most consistent indications, actions to further abate diesel emissions should be prioritized as part of policy measures for protection of public health. However, our results in respect to CO also suggest that there should also be stricter control of emissions from petrol combustion. The role of non-exhaust sources remains a concern however and more extensive monitoring of traffic pollution in urban centres is required to further elucidate the associations.

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The authors declare that they have no actual or potential competing financial interests.

REFERENCES

1. WHO. Review of evidence on health aspects of air pollution – REVIHAAP Project. Technical Report. Copenhagen, Denmark: WHO Publications, WHO Regional Office for Europe; 2013.
2. HEI Panel on the Health Effects of Traffic-Related Air Pollution (2010). Traffic-related air pollution: a critical review of the literature on emissions, exposure, and health effects. Boston, Health Effects Institute Special Report 17. <http://pubs.healtheffects.org/getfile.php?u=553>. Accessed November 28, 2014.
3. Beelen R, Hoek G, van den Brandt PA, et al. Long-term effects of traffic-related air pollution on mortality in a Dutch cohort (NLCS-AIR study). *Environ Health Perspect* 2008;116:196–202.
4. Hoffmann B, Moebus S, Möhlenkamp S, et al. Residential exposure to traffic is associated with coronary atherosclerosis. *Circulation* 2007;116:489–96.
5. Gauderman WJ, Vora H, McConnell R, et al. Effect of exposure to traffic on lung development from 10 to 18 years of age: a cohort study. *Lancet* 2007;369:571-77.
6. Beelen R, Raaschou-Nielsen O, Stafoggia M, et al. Effects of long-term exposure to air pollution on natural-cause mortality: an analysis of 22 European cohorts within the multicentre ESCAPE project. *Lancet* 2014;383:785-95.
7. Raaschou-Nielsen O, Andersen ZJ, Beelen R, et al. Air pollution and lung cancer incidence in 17 European cohorts: prospective analyses from the European Study of Cohorts for Air Pollution Effects (ESCAPE). *Lancet Oncol* 2013;14:813-22.

8. Stafoggia M, Cesaroni G, Peters A, et al. Long-term exposure to ambient air pollution and incidence of cerebrovascular events: results from 11 European cohorts within the ESCAPE project. *Environ Health Perspect* 2014;122:919-25.
9. Cesaroni G, Forastiere F, Stafoggia M, et al. Long term exposure to ambient air pollution and incidence of acute coronary events: prospective cohort study and meta-analysis in 11 European cohorts from the ESCAPE Project. *BMJ* 2014;348:f7412.
10. Gehring U, Gruzieva O, Agius RM, et al. Air pollution exposure and lung function in children: the ESCAPE project. *Environ Health Perspect* 2013;121:1357-64.
11. Fuks KB, Weinmayr G, Foraster M, et al. Arterial blood pressure and long-term exposure to traffic-related air pollution: an analysis in the European Study of Cohorts for Air Pollution Effects (ESCAPE). *Environ Health Perspect* 2014;122:896-905.
12. Peters A, von Klot S, Heier M, et al. Exposure to traffic and the onset of myocardial infarction. *N Engl J Med* 2004;351:1721-30.
13. Zanobetti A, Franklin M, Koutrakis P, et al. Fine particulate air pollution and its components in association with cause-specific emergency admissions. *Environ Health* 2009;8:58.
14. Stafoggia M, Samoli E, Alessandrini E, et al. Short-term associations between fine and coarse particulate matter and hospitalizations in Southern Europe: results from the MED-PARTICLES project. *Environ Health Perspect* 2013;121:1026-33.
15. Samoli E, Aga E, Touloumi G, et al. Short-term effects of nitrogen dioxide on mortality: an analysis within the APHEA project. *Eur Respir J* 2006 ;27:1129-38.

16. Janssen NA, Hoek G, Simic-Lawson M, et al. Black carbon as an additional indicator of the adverse health effects of airborne particles compared with PM10 and PM2.5. *Environ Health Perspect* 2011;119:1691–99.
17. Anderson HR, Armstrong B, Hajat S, et al. Air pollution and activation of implantable cardioverter defibrillators in London. *Epidemiology* 2010;21:405-13.
18. Atkinson RW, Fuller GW, Anderson HR, et al. Urban ambient particle metrics and health: a time-series analysis. *Epidemiology* 2010;21:501-11.
19. Bohnenstenge S, Belcher SE, Aiken A, et al. Meteorology, air quality, and health in London: The ClearfLo project. *Bulletin of the American Meteorological Society*. 2014. ISSN 1520-0477 doi:10.1175/BAMS-D-12-00245.1.
20. Lenschow P, Abraham HJ, Kutzner K, et al. Some ideas about the sources of PM10. *Atmos Environ* 2001;35:S23 – S33.
21. Greater London Authority (GLA). 2012. The London Atmospheric Emissions Inventory available at: <http://datalondongovuk/dataset/london-atmospheric-emissions-inventory-2010> accessed 19th December 2012.
22. Rhys-Tyler GA, Legassick W, Bell MC. The significance of vehicle emissions standards for levels of exhaust pollution from light vehicles in an urban area. *Atmos Environ* 2011;45:3286-93.
23. European Environment Agency (EEA). *Status of black carbon monitoring in ambient air in Europe*. EEA technical report 18/2013. Luxemburg: European Environment Agency; 2013.

24. Thorpe A, Harrison RM. Sources and properties of non-exhaust particulate matter from road traffic: a review. *STOTEN* 2008;400:270-82.
25. Harrison RM, Jones AM, Gietl J, et al. Estimation of the contributions of brake dust, tire wear, and resuspension to nonexhaust traffic particles derived from atmospheric measurements. *Environ Sci Tech* 2012;46:6523-29.
26. Wood SN. Modelling and smoothing parameter estimation with multiple quadratic penalties. *J Royal Stat Soc B* 2000;62:413-28.
27. Mostofsky E, Schwartz J, Coull BA, et al. Modeling the association between particle constituents of air pollution and health outcomes. *Am J Epidemiol* 2012;176:317-26.
28. Atkinson RW, Mills IC, Walton HA, et al.. Fine particle components and health-a systematic review and meta-analysis of epidemiological time series studies of daily mortality and hospital admissions. *J Expo Sci Environ Epidemiol* 2015;25:208-14.
29. Kubesch N, De Nazelle A Guerra S, et al. Arterial blood pressure responses to short-term exposure to low and high traffic-related air pollution with and without moderate physical activity. *Eur J Prev Cardiol* 2015;22:548-57.
30. Krishnan RM, Sullivan JH, Carlsten C, et al. A randomized cross-over study of inhalation of diesel exhaust, hematological indices, and endothelial markers in humans. *Part Fibre Toxicol* 2013;10:7.
31. Anderson HR, Atkinson RW, Bremner SA, et al. Quantitative systematic review of short term associations between ambient air pollution (particulate matter, ozone, nitrogen dioxide, sulphur dioxide and carbon monoxide), and mortality and morbidity. London, Department of Health. 2007.
<https://www.gov.uk/government/publications/quantitative-systematic-review-of-short-term->

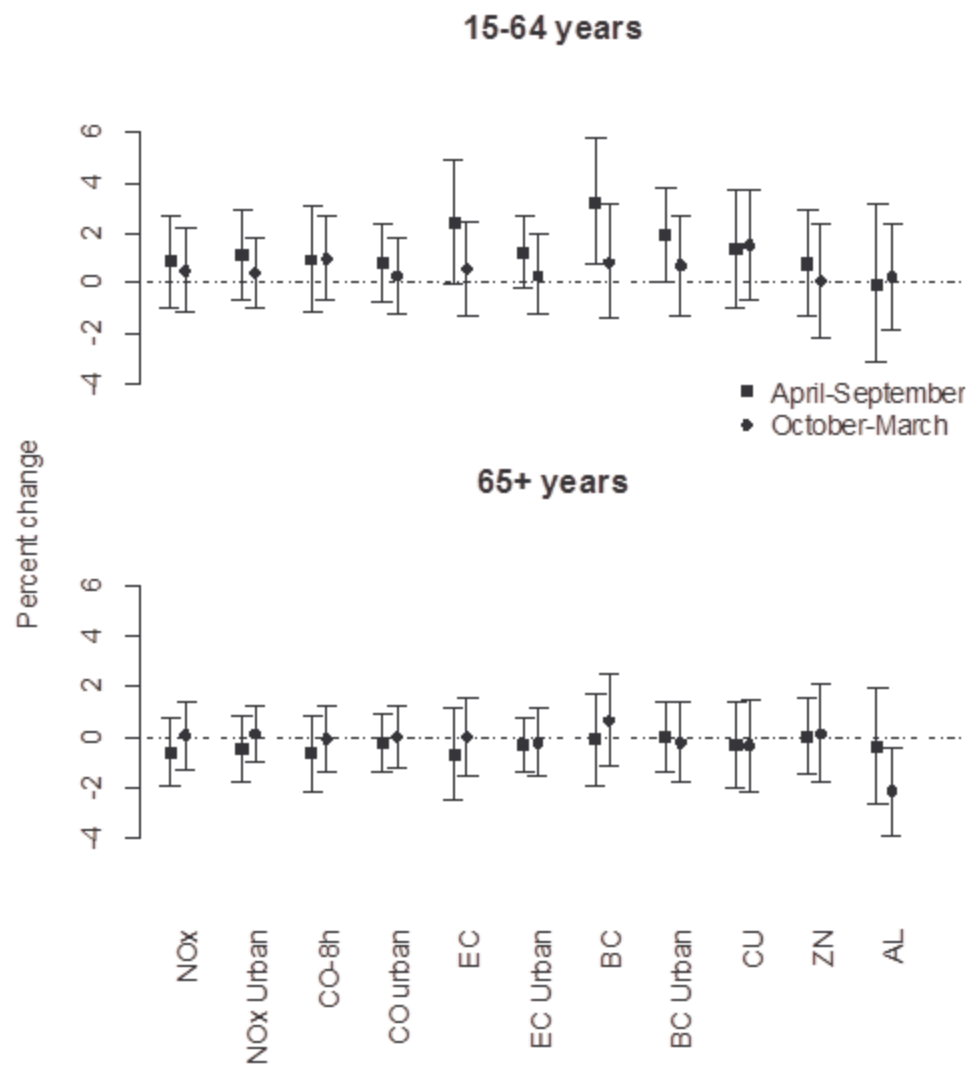
[associations-between-ambient-air-pollution-particulate-matter-ozone-nitrogen-dioxide-sulphur-dioxide-and-carbon-monoxide-and-mortality-and-morbidity](#). Accessed December 4, 2014.

32. Bell ML, Peng RD, Dominici F, et al. Emergency hospital admissions for cardiovascular diseases and ambient levels of carbon monoxide: results for 126 United States urban counties, 1999-2005. *Circulation* 2009;120:949-55.
33. Iskandar A, Andersen ZJ, Bønnelykke K, et al. Coarse and fine, but not ultrafine particles in urban air trigger asthma hospitalizations in children. *Thorax* 2012; 67:252-7.
34. Bell ML, Ebisu K, Leaderer BP, et al. Associations of PM_{2.5} constituents and sources with hospital admissions: analysis of four counties in Connecticut and Massachusetts (USA) for persons ≥ 65 years of age. *Environ Health Perspect* 2014;122:138-44.
35. Basagaña X, Jacquemin B, Karanasiou A, et al. Short-term effects of particulate matter constituents on daily hospitalizations and mortality in five South-European cities: Results from the MED-PARTICLES project. *Environ Intern* 2015;75:151–58.
36. Lippmann M, Chen LC, Gordon T, et al. National Particle Component Toxicity (NPACT) Initiative: integrated epidemiologic and toxicologic studies of the health effects of particulate matter components. *Res Rep Health Eff Inst* 2013;177: 5–13.
37. Le Tertre A, Medina S, Samoli E, et al. Short-term effects of particulate air pollution on cardiovascular diseases in eight European cities. *J Epidemiol Community Health* 2002;56:773-79.

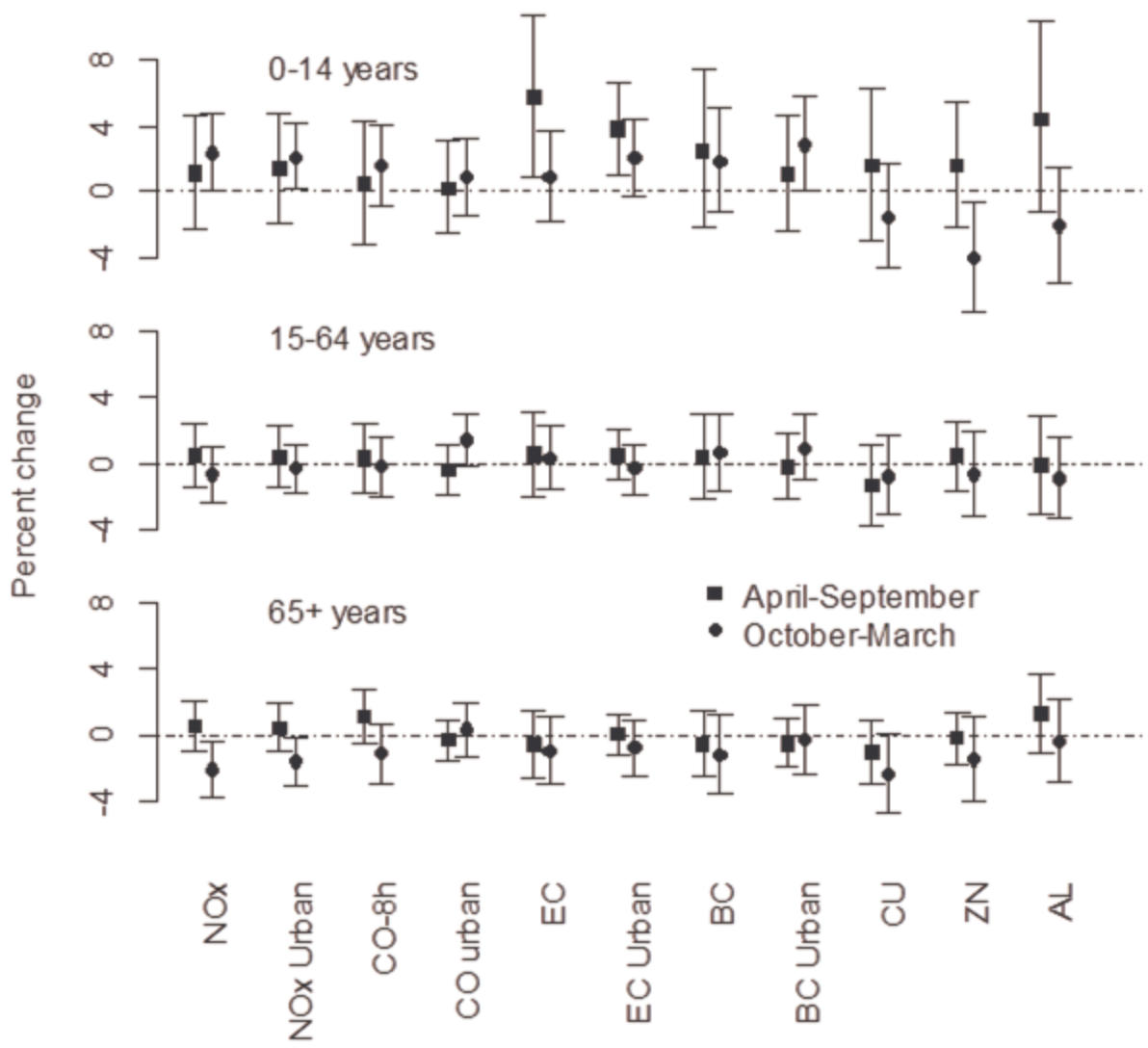
38. Atkinson RW, Anderson HR, Sunyer J, et al. Acute effects of particulate air pollution on respiratory admissions: results from APHEA 2 project. *Air Pollution and Health: a European Approach. Am J Respir Crit Care Med* 2001;164:1860-66.
39. Katsouyanni K, Samet JM, Anderson HR, et al. Air pollution and health: a European and North American approach (APHENA). *Res Rep Health Eff Inst* 2009;142:5-90.
40. Bell ML, Ebisu K, Peng RD, et al. Hospital admissions and chemical composition of fine particle air pollution. *Am J Respir Crit Care Med* 2009;179:1115-20.

Figure 1. Percent change (and 95% confidence intervals (CIs)) in cardiovascular (A) and respiratory (B) hospital admissions associated with an interquartile range increase in traffic pollutants after single day exposure (lag 1 for cardiovascular and lag 2 for respiratory diagnoses) in London, U.K. for 2011–12 during the warm (April to September) and cool (October to March) period of the year. EC/BC and metals are adjusted for particle mass.

(A) Cardiovascular Admissions



(B) Respiratory Admissions



SUPPLEMENTAL MATERIAL

Associations of short-term exposure to traffic-related air pollution with hospital admissions in London, U.K.

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Annex 1. Selection of traffic-related pollution metrics

The use of indicator species to identify emissions from an air pollution source is well established in receptor analysis and source apportionment. Many studies have used enrichment factors of specific indicator species to identify and quantify sources¹ and indicator species are also used in source apportionment to attribute factors to source types. Viana et al.² reviewed PM source apportionment studies conducted across Europe and identified the pollutant species most frequently used in source attribution. These have been used to select indicators of different traffic pollution source mechanisms, along with source specific studies as summarized in Table 1.

Despite being strongly emitted from traffic sources an indicator species might also have other sources in the urban environment. The specificity of the each indicator species will therefore be different. The different transport sources, processes and indicator species are not represented individually in the London Atmospheric Emissions Inventory. We have therefore calculated a kerbside enrichment factor as an index of specificity for London for each tracer during the study period. This has been defined as:

$$\text{Kerbside enrichment factor} = ([\text{Marylebone Road}] - [\text{North Kensington}]) / [\text{North Kensington}]$$

Enrichment factors have been calculated separately for the warm and cool season to reflect potential differing seasonality between the traffic and non-traffic source types.

Table 1. Source indicators and their kerbside enrichment factors.

Source	Indicator	Background	Kerbside enrichment factor
Traffic - general	NOx	NOx, the sum of NO and NO ₂ , is found in greatest concentrations in London close to busy roads. Real-world measurements of exhaust from 72,000 vehicles show greatest NOx emissions arise from diesel and older (pre-EURO 3) petrol vehicles. ³ The London Atmospheric Emissions Inventory shows road transport to be the largest single NOx source in London at 47% of 2010 emissions followed by space heating (16%). ⁴	4.6
Exhaust from petrol vehicles	CO	CO is emitted from incomplete fuel combustion. Real-world vehicle emissions measurements in London ⁵ shows exhaust CO between 1.9% for pre-euro petrol cars to 0.07% for Euro 4. By contrast all diesel vehicle types measured had emissions less than 0.07% and some as low as 0.01%.	1.4
Diesel exhaust	Black and elemental carbon	The black carbon measurement is a function of the light absorption of particles; which is strongly related to the carbon content of the aerosol. Elemental carbon defines the carbon concentration in particles that is not chemically bound. ⁶ Europe-wide these are mainly emitted from transport, especially diesel vehicles. ⁷ Viana et al. ² list the use of black carbon as a tracer for vehicle exhaust in source apportionment studies. Measurements of	5.6 (for BC)

Brake wear particles	Cu	<p>real-world vehicle emissions in London using “smoke number” show that diesel vehicles are overwhelmingly the largest emitters with mean smoke numbers from light duty diesels being around 3.5 times greater than those from petrol.⁵ Viana et al.² list the use of Cu as an indicator of traffic emissions from brake wear alongside Ba and Sb. The chemical composition of brake linings and brake dust vary according to product and application but due to its use as a high temperature lubricant Cu is generally the most abundant element in brake linings and is found in high abundance in brake dust.⁸</p>	4.7
Tyre wear particles	Zn	<p>Tyres are around 1% Zn by weight. It is used as an activator in the vulcanization process and is the only element in tyres that are present at significantly greater than crustal abundance.⁸ Viana et al.² list the use of Zn as indicator of traffic emissions from tyre wear. Zn has also been used as an indicator of emissions from lubricating oil however traffic emissions are dominated (>90%) by tyre wear sources.⁹ This was supported by Harrison et al.⁹ who found negligible concentrations of sub-micron Zn in the roadside increment in London.</p>	1.3
Mineral dust	Al	<p>Viana et al.² list the use of Al, Si, Ca and Fe in PM₁₀ as indicators of crustal / mineral particles. Frank¹⁰ used an equation using Si, Ca, Fe and Ti to apportion crustal material. However in urban settings Fe might also originate from vehicle wear sources for instance Harrison et al.⁹ used Fe measured in the PM coarse as a marker for vehicle and soil dust in London. Of the remaining metallic elements measured in this study, Al and Ca occur in sufficient quantities to be used as tracers for mineral dusts. Given the identification of Ca in lubricating oil emissions from traffic in London,¹¹ Al was selected as the favored indicator species.</p>	1.3

References

1. Watson JG, Chow JC, Lowenthal DH, Pritchett LC, Frazier CA, Neuroth GR et al. Differences in the carbon composition of source profiles for diesel-and gasoline-powered vehicles. *Atmos Environ.* 1994; 28:2493-2505.
2. Viana M, Kuhlbusch TAJ, Querol X, Alastuey A, Harrison RM, Hopke PK et al. Source apportionment of particulate matter in Europe: a review of methods and results. *J Aer Scien.* 2008; 39: 827-849.
3. Carslaw DC, Rhys-Tyler G. New insights from comprehensive on-road measurements of NO_x, NO₂ and NH₃ from vehicle emission remote sensing in London, UK. *Atmos Environ* 2013; 81:339-347.
4. Greater London Authority (GLA). 2012. The London Atmospheric Emissions Inventory available at: <http://datalondongovuk/dataset/london-atmospheric-emissions-inventory-2010> accessed 19th December 2012.
5. Rhys-Tyler GA, Legassick W, Bell MC. The significance of vehicle emissions standards for levels of exhaust pollution from light vehicles in an urban area. *Atmos Environ.* 2011;45:3286-3293.
6. Petzold A, Ogren JA, Fiebig M, Laj P, Li SM, Baltensperger U et al. Recommendations for reporting “black carbon” measurements. *Atmos Chem Phys.* 2013; 13: 8365–8379.
7. European Environment Agency (EEA). *Status of black carbon monitoring in ambient air in Europe. EEA technical report 18/2013.* Luxemburg: European Environment Agency; 2013.
8. Thorpe A, Harrison RM. 2008. Sources and properties of non-exhaust particulate matter from road traffic: a review. *STOTEN* 2008;400:270-282.

9. Harrison RM, Jones AM, Gietl J, Yin J, Green DC. Estimation of the contributions of brake dust, tire wear, and resuspension to nonexhaust traffic particles derived from atmospheric measurements. *Environ Sci Tech.* 2012;46:6523-6529.
10. Frank NH. Retained nitrate, hydrated sulfates, and carbonaceous mass in federal reference method fine particulate matter for six eastern US cities. *J Air Waste Manag Ass*, 2006;56: 500-511.
11. Dall'Osto M, Harrison RM. Urban organic aerosols measured by single particle mass spectrometry in the megacity of London. *Atmos Chem Phys.* ,2012;12:4127-4142.

Annex 2. Urban increment estimation

Source apportionment of several metrics was undertaken using the approach first proposed by Lenschow et al.¹ This approach assumes that the measured concentrations in urban areas consist of the sum of contributions from three different source areas, namely:

The regional background: This is the concentration of the pollutant present in air around the city. It is assumed to vary in time, but not space over the city and its surroundings. In the TRAFFIC study measurement sites to the west (Harwell, Oxfordshire), east (Detling, Kent) and, for CO only Egham in Surrey, were used to determine regional concentrations outside the London plume.

The urban background: This concentration is from the sum of all urban sources. The urban increment (urban – regional) is the concentration from urban sources. The urban increment for NO_x, CO and BC were included in the analysis as specific tracers for urban traffic sources, as distinct from contributions from industry and more distant sources that determine the regional background. In this study the North Kensington measurement site was used.

The kerb / roadside: This is the concentration from a road source when measured very nearby. The contribution from the nearby road can be deduced from the difference between traffic and urban background monitoring sites. In this study the Marylebone Road measurement site was used to calculate kerbside enrichment factors.

References

1. Lenschow P, Abraham HJ, Kutzner K, Lutz M, Preuß JD, Reichenbächer W. Some ideas about the sources of PM₁₀ *Atmos Environ.* 2001;35: S23-S33.

Annex 3. Supplemental description of the selected pollutants

Table 1. Distribution of the pollutants stratified by warm (April to September) and cool (October to March) period of the year.

Pollutant ($\mu\text{g}/\text{m}^3$, except CO mg/m^3)	April to September				October to March			
	Mean	Median	IQR ^a	90th percentile	Mean	Median	IQR ^a	90th percentile
NO_x	37.4	31.1	20	63.1	72.9	59.7	55.0	139.0
NO_x - Urban increment	28.0	21.8	17.1	49.6	56.6	43.9	41.2	136.3
CO	0.3	0.3	0.1	0.4	0.4	0.4	0.2	0.6
CO –Urban Increment	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.2
EC	0.8	0.6	0.6	1.4	1.3	1.1	0.9	2.4
EC - Urban	0.6	0.5	0.3	1.1	0.9	0.7	0.6	1.8
BC	1.2	1.0	0.7	2.0	1.8	1.4	1.3	3.4
BC –Urban	0.8	0.6	0.4	1.4	1.1	0.9	0.9	2.1
Cu	0.0074	0.0063	0.01	0.142	0.0114	0.0087	0.01	0.0228
Zn	0.0093	0.0071	0.01	0.060	0.0150	0.0110	0.01	0.0318
Al	0.0765	0.0552	0.07	0.1555	0.0751	0.0567	0.06	0.1458
PM₁₀	15.2	13.1	8.0	24.0	21.5	18.0	12.5	38.0
PM_{2.5}	8.9	7.3	4.1	14.8	15.5	11.0	12.1	32.0
NO₂	28.9	26.1	14.5	45.2	43.5	43.0	22.9	62.9
SO₂	1.60	1.7	2.4	3.0	2.1	2.0	2.1	4.1
O₃	70.1	67.2	26.9	97.6	41.0	44.0	26.3	63.7

^aIQR: Interquartile Range

Table 2. Annual correlations between pollutants.

Pollutant	NOx	NOx-Urban	CO	CO-Urban	EC	EC-Urban	BC	BC-Urban	C	Zn	Al	PM₁₀	PM_{2.5}	NO₂	SO₂
NOx	1														
NOx – Urban	0.98	1													
CO	0.83	0.81	1												
CO –Urban	0.35	0.41	0.60	1											
EC	0.91	0.90	0.74	0.37	1										
EC - Urban	0.78	0.83	0.62	0.45	0.92	1									
BC	0.90	0.88	0.77	0.36	0.92	0.78	1								
BC –Urban	0.81	0.83	0.66	0.35	0.86	0.85	0.92	1							
Cu	0.77	0.76	0.62	0.32	0.81	0.69	0.78	0.65	1						
Zn	0.68	0.63	0.57	0.13	0.68	0.45	0.68	0.45	0.74	1					
Al	0.36	0.33	0.26	0.00	0.40	0.28	0.40	0.26	0.49	0.55	1				
PM₁₀	0.65	0.57	0.57	0.04	0.53	0.32	0.60	0.38	0.56	0.71	0.65	1			
PM_{2.5}	0.65	0.57	0.58	0.05	0.54	0.33	0.61	0.39	0.56	0.71	0.55	0.95	1		
NO₂	0.90	0.87	0.74	0.23	0.83	0.69	0.80	0.72	0.71	0.66	0.44	0.66	0.66	1	
SO₂	0.55	0.51	0.42	0.08	0.46	0.24	0.48	0.32	0.40	0.47	0.37	0.51	0.49	0.52	1
O₃	-0.48	-0.45	-0.42	-0.13	-0.36	-0.23*	-0.39	-0.39	-0.27	-0.23	0.19	-0.19	-0.28	-0.40	-0.17

Annex 4. Percent change (and 95% confidence intervals (CIs)) in cardiovascular and respiratory hospital admissions associated with an interquartile range increase (in $\mu\text{g}/\text{m}^3$) in regulated pollutants following single day (lag 1 for cardiovascular and lag 2 for respiratory diagnoses) or weekly exposure (lags 0-6) in London, U.K. for 2011–12.

Pollutants	CVD Admissions % (95% CI)		Respiratory Admissions % (95% CI)		
	15-64 years	65+ years	0-14 years	15-64 years	65+ years
Single Day Exposure					
PM₁₀	0.17 (-0.86, 1.21)	-0.50 (-1.27, 0.28)	0.69 (-0.85, 2.25)	-0.67 (-1.69, 0.37)	-1.14 (-2.10, -0.16)
PM_{2.5}	0.19 (-0.73, 1.12)	-0.80 (-1.49, -0.10)	0.42 (-0.94, 1.81)	-0.80 (-1.72, 0.13)	-0.97 (-1.84, -0.09)
NO₂	1.00 (-0.87, 2.91)	-0.75 (-2.15, 0.68)	1.91 (-0.78, 4.67)	-1.14 (-2.94, 0.69)	-3.11 (-4.75, -1.44)
SO₂	0.79 (-0.72, 2.32)	-0.15 (-1.29, 1.00)	0.69 (-1.38, 2.81)	-0.90 (-2.38, 0.61)	-1.93 (-3.26, -0.57)
O₃	-0.12 (-2.01, 1.81)	1.58 (0.13, 3.05)	-1.64 (-4.35, 1.15)	3.23 (1.31, 5.19)	4.83 (3.01, 6.68)
Weekly Exposure					
PM₁₀	-0.76 (-2.32, 0.81)	-0.32 (-1.50, 0.87)	2.59 (-0.24, 5.50)	-0.61 (-2.17, 0.98)	-3.28 (-4.80, -1.73)
PM_{2.5}	-0.63 (-2.07, 0.83)	-0.66 (-1.74, 0.43)	2.21 (-0.33, 4.81)	-0.92 (-2.36, 0.54)	-3.03 (-4.43, -1.60)
NO₂	-1.52 (-4.47, 1.53)	-0.16 (-2.42, 2.16)	6.30 (1.07, 11.81)	-1.69 (-4.65, 1.35)	-7.16 (-9.89, -4.35)
SO₂	-1.32 (-4.07, 1.51)	-0.35 (-2.44, 1.79)	4.95 (0.54, 9.56)	-1.99 (-4.75, 0.87)	-3.21 (-5.77, -0.58)
O₃	0.66 (-2.22, 3.63)	1.95 (-0.23, 4.19)	-7.89 (-12.39, -3.16)	4.71 (1.66, 7.85)	8.84 (5.73, 12.04)

Annex 5. Results from multi pollutants’ models. Percent change (and 95% confidence intervals (CIs)) in cardiovascular and respiratory hospital admissions associated with an interquartile range increase in traffic pollutants after single day exposure (lag 1 for cardiovascular and lag 2 for respiratory diagnoses) in London, U.K. for 2011–12. EC/BC and metals are also adjusted for PM mass.

Indicator/ Pollutant	Controlling for	Cardiovascular Admissions % (95% CI)		Respiratory Admissions % (95% CI)		
		15-64 years	65+ years	0-14 years	15-64 years	65+ years
General traffic						
<i>NO_x</i>	<i>O₃</i>	1.05 (-0.27, 2.38)	0.24 (-0.76, 1.25)	0.74 (-1.01, 2.52)	0.12 (-1.14, 1.40)	-0.59 (-1.74, 0.57)
	<i>SO₂</i>	0.74 (-0.65, 2.16)	-0.38 (-1.44, 0.69)	0.96 (-0.85, 2.80)	-0.62 (-1.99, 0.77)	-1.42 (-2.66, -0.16)
	<i>PM_{2.5}</i>	1.22 (-0.26, 2.73)	0.54 (-0.59, 1.69)	0.98 (-0.88, 2.88)	-0.31 (-1.73, 1.13)	-1.65 (-2.92, -0.36)
<i>NO_x Urban</i>	<i>O₃</i>	1.07 (-0.13, 2.29)	0.36 (-0.56, 1.28)	0.82 (-0.75, 2.41)	0.22 (-0.93, 1.39)	-0.47 (-1.53, 0.59)
	<i>SO₂</i>	0.77 (-0.48, 2.05)	-0.14 (-1.10, 0.83)	0.94 (-0.68, 2.58)	-0.42 (-1.65, 0.84)	-1.12 (-2.25, 0.02)
	<i>PM_{2.5}</i>	1.12 (-0.17, 2.41)	0.53 (-0.45, 1.52)	0.97 (-0.64, 2.61)	-0.17 (-1.40, 1.08)	-1.27 (-2.38, -0.14)
	<i>EC Urban</i>	-2.02 (-4.79, 0.84)	0.91 (-1.33, 3.20)	-0.24 (-4.03, 3.70)	-2.78 (-5.58, 0.10)	-2.12 (-4.68, 0.52)
Petrol vehicle exhaust						
<i>CO</i>	<i>O₃</i>	1.86 (0.29, 3.46)	-0.22 (-1.42, 0.98)	0.79 (-1.39, 3.01)	-0.37 (-1.93, 1.22)	-0.85 (-2.28, 0.61)
	<i>SO₂</i>	1.70 (0.05, 3.39)	-0.62 (-1.88, 0.65)	0.70 (-1.54, 3.00)	-1.12 (-2.77, 0.56)	-1.69 (-3.20, -0.15)

	<i>PM_{2.5}</i>	2.09 (0.31, 3.91)	0.19 (-1.17, 1.56)	0.95 (-1.40, 3.35)	-0.61 (-2.37, 1.17)	-1.80 (-3.38, -0.20)
<i>CO Urban</i>	<i>O₃</i>	0.96 (-0.07, 2.00)	-0.07 (-0.85, 0.72)	0.94 (-0.44, 2.35)	0.37 (-0.67, 1.41)	-0.32 (-1.25, 0.62)
	<i>SO₂</i>	1.02 (-0.03, 2.07)	-0.20 (-1.00, 0.60)	0.81 (-0.60, 2.24)	0.17 (-0.88, 1.23)	-0.48 (-1.43, 0.49)
	<i>PM_{2.5}</i>	0.96 (-0.07, 1.99)	-0.06 (-0.84, 0.72)	0.91 (-0.47, 2.31)	0.29 (-0.74, 1.32)	-0.49 (-1.42, 0.46)
	<i>EC Urban</i>	0.43 (-0.87, 1.75)	0.02 (-0.99, 1.03)	0.38 (-1.38, 2.17)	0.26 (-1.04, 1.59)	-0.60 (-1.77, 0.58)
Diesel vehicle exhaust						
<i>EC</i>	<i>NO_x</i>	4.15 (0.88, 7.53)	0.54 (-1.91, 3.05)	-0.42 (-4.87, 4.24)	0.13 (-3.03, 3.39)	1.16 (-1.85, 4.27)
	<i>CO</i>	0.90 (-1.05, 2.90)	0.38 (-1.11, 1.90)	0.79 (-1.82, 3.47)	0.32 (-1.62, 2.30)	-0.17 (-1.96, 1.65)
<i>EC Urban</i>	<i>NO_x</i>	2.49 (0.57, 4.45)	0.05 (-1.42, 1.55)	1.65 (-1.08, 4.45)	2.09 (0.17, 4.06)	1.63 (-0.15, 3.45)
	<i>CO</i>	1.00 (-0.37, 2.38)	0.21 (-0.84, 1.28)	0.97 (-0.87, 2.85)	0.73 (-0.63, 2.12)	0.72 (-0.53, 1.99)
<i>BC</i>	<i>NO_x</i>	3.32 (-0.22, 6.98)	0.33 (-2.31, 3.03)	-1.32 (-6.18, 3.79)	-0.60 (-4.01, 2.92)	2.90 (-0.45, 6.37)
	<i>CO</i>	0.40 (-1.74, 2.59)	0.74 (-0.91, 2.41)	0.68 (-2.14, 3.59)	0.65 (-1.51, 2.86)	-0.01 (-2.02, 2.03)
<i>BC Urban</i>	<i>NO_x</i>	0.55 (-1.66, 2.82)	-0.63 (-2.31, 1.08)	0.59 (-2.48, 3.76)	0.28 (-1.91, 2.52)	1.85 (-0.19, 3.94)
	<i>CO</i>	-0.21 (-1.74, 1.35)	0.00 (-1.18, 1.19)	0.64 (-1.40, 2.71)	0.76 (-0.78, 2.32)	0.79 (-0.60, 2.21)
Vehicle Non-exhaust						
<i>Cu</i>	<i>NO_x</i>	1.37 (-0.56, 3.34)	0.03 (-1.45, 1.53)	-1.40 (-3.97, 1.25)	-1.55 (-3.45, 0.38)	-1.07 (-2.84, 0.74)
	<i>CO</i>	0.80 (-0.85, 2.47)	0.15 (-1.11, 1.42)	-0.27 (-2.47, 1.99)	-1.16 (-2.80, 0.51)	-1.31 (-2.82, 0.23)

<i>ZN</i>	<i>NOx</i>	-0.27 (-1.73, 1.22)	0.17 (-0.96, 1.31)	-1.85 (-3.85, 0.20)	-0.44 (-1.94, 1.09)	-0.14 (-1.51, 1.24)
	<i>CO</i>	-0.38 (-1.76, 1.03)	0.21 (-0.85, 1.28)	-1.16 (-3.05, 0.77)	-0.18 (-1.6, 1.26)	-0.40 (-1.69, 0.91)
<i>Al</i>	<i>NOx</i>	0.55 (-1.08, 2.21)	-1.13 (-2.37, 0.12)	-0.39 (-2.84, 2.11)	0.79 (-0.89, 2.50)	1.62 (0.06, 3.21)
	<i>CO</i>	0.39 (-1.22, 2.03)	-1.13 (-2.35, 0.10)	0.20 (-2.22, 2.68)	0.86 (-0.79, 2.55)	1.45 (-0.09, 3.01)

Annex 6. Percent change (and 95% confidence intervals (CIs)) in hospital admissions by age group and season associated with season-specific interquartile range (IQR) increase in traffic-related pollutants after single day exposure (lag 1 for cardiovascular (A) and lag 2 for respiratory (B) diagnoses) in London, U.K. for 2011–12. EC/BC and metals are adjusted for PM mass.

(A) Cardiovascular Admissions

Indicator/Pollutants	15-64 years % (95% CI)		65+ years % (95% CI)	
	April-September	October-March	April-September	October-March
General traffic				
<i>NOx</i>	0.87 (-0.96, 2.73)	0.53 (-1.09, 2.17)	-0.59 (-1.94, 0.78)	0.07 (-1.24, 1.40)
<i>NOx – Urban</i>	1.12 (-0.62, 2.89)	0.41 (-0.99, 1.84)	-0.40 (-1.69, 0.90)	0.13 (-1.01, 1.29)
Petrol vehicle exhaust				
<i>CO</i>	0.95 (-1.08, 3.02)	0.96 (-0.67, 2.61)	-0.63 (-2.14, 0.90)	-0.07 (-1.39, 1.27)
<i>CO –Urban</i>	0.82 (-0.69, 2.36)	0.31 (-1.20, 1.84)	-0.22 (-1.34, 0.92)	0.02 (-1.21, 1.26)
Diesel vehicle exhaust				
<i>EC</i>	2.41 (-0.03, 4.90)	0.56 (-1.30, 2.46)	-0.66 (-2.47, 1.18)	0.00 (-1.51, 1.54)
<i>EC - Urban</i>	1.23 (-0.16, 2.63)	0.33 (-1.25, 1.94)	-0.28 (-1.33, 0.78)	-0.21 (-1.53, 1.13)
<i>BC</i>	3.22 (0.75, 5.74)	0.85 (-1.37, 3.11)	-0.10 (-1.91, 1.74)	0.66 (-1.14, 2.49)
<i>BC –Urban</i>	1.92 (0.08, 3.80)	0.68 (-1.32, 2.73)	0.04 (-1.34, 1.44)	-0.19 (-1.78, 1.43)

Vehicle Non-exhaust

<i>Cu</i>	1.33 (-0.99, 3.70)	1.51 (-0.66, 3.73)	-0.28 (-1.97, 1.45)	-0.32 (-2.13, 1.51)
<i>Zn</i>	0.77 (-1.30, 2.87)	0.07 (-2.17, 2.37)	0.06 (-1.44, 1.58)	0.14 (-1.76, 2.07)
<i>Al</i>	-0.06 (-3.13, 3.11)	0.23 (-1.85, 2.36)	-0.37 (-2.65, 1.97)	-2.12 (-3.84, -0.38)

(B) Respiratory Admissions

	0-14 years % (95%CI)		15-64 years % (95%CI)		65+ years % (95%CI)	
	April-September	October-March	April-September	October-March	April-September	October-March
General traffic						
<i>NOx</i>	1.14 (-2.26, 4.67)	2.40 (0.09, 4.76)	0.47 (-1.40, 2.38)	-0.65 (-2.30, 1.02)	0.57 (-0.93, 2.08)	-2.09 (-3.75, -0.39)
<i>NOx – Urban</i>	1.41 (-1.89, 4.81)	2.14 (0.14, 4.17)	0.40 (-1.41, 2.23)	-0.32 (-1.75, 1.13)	0.46 (-0.98, 1.92)	-1.60 (-3.06, -0.12)
Petrol vehicle exhaust						
<i>CO</i>	0.52 (-3.18, 4.36)	1.62 (-0.82, 4.12)	0.33 (-1.71, 2.42)	-0.20 (-1.93, 1.55)	1.19 (-0.44, 2.85)	-1.13 (-2.89, 0.67)
<i>CO –Urban</i>	0.24 (-2.49, 3.05)	0.84 (-1.43, 3.16)	-0.37 (-1.87, 1.16)	1.41 (-0.19, 3.03)	-0.31 (-1.52, 0.90)	0.33 (-1.32, 2.01)
Diesel vehicle exhaust						
<i>EC</i>	5.67 (0.92, 10.64)	0.93 (-1.80, 3.73)	0.58 (-1.94, 3.16)	0.39 (-1.50, 2.31)	-0.57 (-2.58, 1.49)	-0.93 (-2.87, 1.05)
<i>EC - Urban</i>	3.81 (1.02, 6.68)	2.10 (-0.19, 4.44)	0.52 (-0.98, 2.05)	-0.31 (-1.81, 1.22)	0.05 (-1.16, 1.27)	-0.76 (-2.42, 0.92)
<i>BC</i>	2.48 (-2.15, 7.33)	1.86 (-1.23, 5.04)	0.41 (-2.09, 2.97)	0.66 (-1.62, 3.00)	-0.51 (-2.48, 1.50)	-1.16 (-3.48, 1.21)
<i>BC –Urban</i>	1.04 (-2.42, 4.63)	2.89 (0.04, 5.83)	-0.18 (-2.06, 1.75)	0.99 (-1.05, 3.07)	-0.47 (-1.91, 0.98)	-0.27 (-2.35, 1.86)

**Vehicle Non-
exhaust**

<i>Cu</i>	1.54 (-2.92, 6.22)	-1.53 (-4.65, 1.70)	-1.31 (-3.72, 1.17)	-0.76 (-3.13, 1.66)	-1.01 (-2.91, 0.92)	-2.26 (-4.64, 0.18)
<i>Zn</i>	1.55 (-2.21, 5.45)	-3.99 (-7.30, -0.56)	0.48 (-1.56, 2.57)	-0.64 (-3.18, 1.97)	-0.21 (-1.80, 1.41)	-1.48 (-4.05, 1.16)
<i>Al</i>	4.39 (-1.17, 10.26)	-2.06 (-5.46, 1.45)	-0.14 (-3.12, 2.94)	-0.90 (-3.34, 1.60)	1.25 (-1.08, 3.63)	-0.30 (-2.80, 2.27)