UC Irvine UC Irvine Previously Published Works

Title

Associations of oxidative stress and inflammatory biomarkers with chemicallycharacterized air pollutant exposures in an elderly cohort.

Permalink https://escholarship.org/uc/item/5r3291m0

Authors

Zhang, Xian Staimer, Norbert Gillen, Daniel L <u>et al.</u>

Publication Date

2016-10-01

DOI

10.1016/j.envres.2016.06.019

Peer reviewed

Contents lists available at ScienceDirect

Environmental Research

ISEVIER



CrossMark

journal homepage: www.elsevier.com/locate/envres

Associations of oxidative stress and inflammatory biomarkers with chemically-characterized air pollutant exposures in an elderly cohort

Xian Zhang^a, Norbert Staimer^a, Daniel L. Gillen^b, Tomas Tjoa^a, James J. Schauer^c, Martin M. Shafer^c, Sina Hasheminassab^d, Payam Pakbin^d, Nosratola D. Vaziri^e, Constantinos Sioutas^d, Ralph J. Delfino^{a,*}

^a Department of Epidemiology, School of Medicine, University of California, Irvine, CA, USA

^b Department of Statistics, School of Information and Computer Sciences, University of California, Irvine, CA, USA

^c University of Wisconsin-Madison, Environmental Chemistry and Technology Program, Madison, WI, USA

^d Department of Civil and Environmental Engineering, Viterbi School of Engineering, University of Southern California, Los Angeles, CA, USA

e Division of Nephrology and Hypertension, Department of Medicine, School of Medicine, University of California, Irvine, CA, USA

ARTICLE INFO

Article history: Received 14 March 2016 Received in revised form 18 May 2016 Accepted 10 June 2016

Keywords: Air pollution Inflammation Oxidative stress Particulate matter components Biomarkers

ABSTRACT

Background: Exposure to air pollution has been associated with cardiorespiratory morbidity and mortality. However, the chemical constituents and pollution sources underlying these associations remain unclear.

Method: We conducted a cohort panel study involving 97 elderly subjects living in the Los Angeles metropolitan area. Airway and circulating biomarkers of oxidative stress and inflammation were measured weekly over 12 weeks and included, exhaled breath condensate malondialdehyde (EBC MDA), fractional exhaled nitric oxide (FeNO), plasma oxidized low-density lipoprotein (oxLDL), and plasma interleukin-6 (IL-6). Exposures included 7-day personal nitrogen oxides (NO_x), daily criteria-pollutant data, five-day average particulate matter (PM) measured in three size-fractions and characterized by chemical components including transition metals, and *in vitro* PM oxidative potential (dithiothreitol and macrophage reactive oxygen species). Associations between biomarkers and pollutants were assessed using linear mixed effects regression models.

Results: We found significant positive associations of airway oxidative stress and inflammation with traffic-related air pollutants, ultrafine particles and transition metals. Positive but nonsignificant associations were observed with PM oxidative potential. The strongest associations were observed among PM variables in the ultrafine range (PM < 0.18 μ m). It was estimated that an interquartile increase in 5-day average ultrafine polycyclic aromatic hydrocarbons was associated with a 6.3% (95% CI: 1.1%, 11.6%) increase in EBC MDA and 6.7% (95% CI: 3.4%, 10.2%) increase in FeNO. In addition, positive but nonsignificant associations were observed between oxLDL and traffic-related pollutants, ultrafine particles and transition metals while plasma IL-6 was positively associated with 1-day average traffic-related pollutants.

Conclusion: Our results suggest that exposure to pollutants with high oxidative potential (traffic-related pollutants, ultrafine particles, and transition metals) may lead to increased airway oxidative stress and inflammation in elderly adults. This observation was less clear with circulating biomarkers.

© 2016 Elsevier Inc. All rights reserved.

1. Introduction

Epidemiological studies have shown positive associations between short-term exposures to air pollutants and cardiopulmonary morbidity and mortality as reviewed by Franklin et al. (2015). The particular mechanisms linking air pollution to acute respiratory and cardiovascular events are not completely understood. Particulate matter (PM) air pollution consists of discrete particles that range from coarse-sized particles $2.5-10 \,\mu\text{m}$ (PM_{2.5-10}), to accumulation mode particles $0.1-2.5 \,\mu\text{m}$ (PM_{0.1-2.5}), and finally to ultrafine particles $< 0.1 \,\mu\text{m}$ in diameter (PM_{0.1}). Particle size is an important determinant of deposition in the respiratory tract, and an indicator of chemical composition and source (Delfino et al., 2005; Sioutas et al., 2005). Smaller particles have a higher pulmonary deposition fraction and penetrate deeper in the lung (Lippmann, 1977). The large surface area of ultrafine



http://dx.doi.org/10.1016/j.envres.2016.06.019 0013-9351/© 2016 Elsevier Inc. All rights reserved.

particles also carries high concentrations of pro-oxidant chemical components, such as polycyclic aromatic hydrocarbons (PAHs) and transition metals, each of which has been shown to induce oxidative stress responses (Li et al., 2003) and can translocate from pulmonary sites to the circulatory system (Elder et al., 2006). Experimental studies have provided evidence that ultrafine particles can induce the greatest amount of oxidative stress and inflammation per unit of PM mass (Li et al., 2002; Cho et al., 2005). This may in-turn impact physiologic responses that ultimately increase the risk of acute cardiorespiratory morbidity (Weichenthal, 2012). In epidemiological studies, some have shown that ultrafine particulate matter air pollution is more strongly associated with adverse health effects when compared to larger PM diameters (Delfino et al., 2009; Franck et al., 2011) while other studies have shown PM2.5 to have stronger or as strong associations as ultrafine PM (Ruckerl et al., 2014; Lanzinger et al., 2016). The inconsistent epidemiological evidence may be due to the fact that the sources and components of pollutants vary from study to study. In urban cities, such as Los Angeles, traffic-related pollutants are the main source for PM mass concentrations (Pant et al., 2013). Those traffic-related pollutants have been shown to contain redox active chemicals that are able to generate reactive oxygen species (ROS) responsible for increases in oxidative stress (Ayres et al., 2008). By quantifying the inherent capacity of PM to oxidize target molecules, oxidative potential is proposed to be a more attractive and biologically more relevant exposure metric than PM mass (Borm et al., 2007). Therefore, the direct measurement of PM oxidative potential may show stronger or more precise associations than either ultrafine PM or traffic-related air pollutants.

Most previous epidemiologic studies have used ambient air monitoring data from central air monitoring stations, and focused primarily on the U.S. Environmental Protection Agency (EPA)'s criteria air pollutants, namely, PM_{2.5}, PM₁₀, ozone (O₃), nitrogen dioxide (NO₂) and carbon monoxide (CO). Several research groups including ours have been investigating the associations between biomarkers of effect and size-fractionated and/or chemicallycharacterized particulate air pollutants (Delfino et al., 2008, 2009, 2010a; Chen et al., 2015; Wu et al., 2015). However, there is still a need to systematically explore oxidative stress and inflammatory biomarkers in both pulmonary and circulatory systems in relation to chemically-characterized PM.

To better understand the underlying mechanism of the association between cardiorespiratory morbidity and short-term exposure to air pollution, we conducted a panel study to investigate the potential roles of air pollutant components and pollution source tracers on both airway and systemic biomarkers of oxidative stress and inflammation in an elderly cohort. The focus on an elderly cohort is motivated by findings that reveal that older subjects tend to be more susceptible to air pollution-induced health effects, and hence represent a high risk population (Brook et al., 2010). We hypothesized that chemicals from fossil fuel combustion, related ultrafine particles, and PM oxidative potential would have stronger adverse health effects than other air pollutant variables. To our knowledge, this is the first study to relate biomarkers of airway and systemic oxidative stress responses to exposure markers of PM oxidative potential in an elderly population.

2. Methods

2.1. Study design

To investigate short-term health effects of exposures to air pollution, we conducted a cohort panel study of repeated measures of outcomes and exposures for 97 elderly non-smoking adults (age \geq 65) living in two Los Angeles California metropolitan

areas (downtown Los Angeles and Anaheim, CA) between 2012 and 2014. The repeated measures design can effectively allow for the control of between and within-subject variability in regression models. A study design flowchart can be found in Appendix, Fig. A.1. Briefly, two groups of subjects in each area were followed alternatively for up to 12 weeks in two discrete 6-week periods in order to incorporate seasonal differences in air pollution levels in the Los Angeles metropolitan area (Daher et al., 2013; Hasheminassab et al., 2014b). Specifically, one 6-week period took place during the warm season (July–October) and the other took place the cool season (November–February). Weekly clinical visits for study participants were scheduled on the same day of week and at the same time of day in order to minimize potential biases induced by weekly and circadian variation (94% of the subjects arrived to clinic with a variability of ± 2 h).

Subjects were excluded from study participation if they lived or were employed outside of the monitored community (18 km radius), smoked within the last 12 months, abused drugs or alcohol, or reported exposure to environmental tobacco smoke at home or on a regular basis at other locations. Additional health criteria for exclusion included the presence of psychiatric disorders or dementia that would prevent the subject from full participation, dialysis treatment or renal failure, daily oral corticosteroids, and active cancer. Observations following the previous 7 days when subjects reported any acute infection (7.62%) were excluded *a priori* given the known major impact of infections on systemic and respiratory inflammation.

2.2. Baseline questionnaires

Background questionnaires collected at the beginning of the study included information on socioeconomic status, medical history, current medication use, history of smoking, and environmental exposure profile. In addition, a baseline fasting blood sample was taken to obtain plasma lipid profiles and glucose levels.

2.3. Markers of airway oxidative stress and inflammation

2.3.1. Exhaled breath condensate (EBC) sampling and malondialdehyde (MDA) analysis

We collected EBC samples during normal breathing with the RTube[™] Collection System (Respiratory Research, Inc., Austin, TX) using standard procedures recommended by the American Thoracic Society and European Respiratory Society (Holvoet et al., 2003). Room air was inhaled through a one-way valve and the exhaled air was directed into a collection chamber (solid aluminum tube pre-chilled on dry ice) where vapors, aerosols and moisture in the breath condense. EBC was then collected from the walls of the condenser. The samples were transported to our laboratory on dry ice and then frozen at -80 °C until analysis. We analyzed EBC samples for MDA using HPLC analysis by modifying the Larstad et al. (2002) protocol (details are given in the online supplement). The estimated limit of quantification (LOQ) for MDA in EBC samples is 3 nM. All values < LOQ were set to 1.5 nmol. We excluded MDA results if the concentration was greater than the upper limit of quantification (12.5 nM) and coefficient of variation (CV) > 25% (occurring in 1.16% of samples).

2.3.2. Fractional concentration of exhaled nitric oxide (FeNO) measurement

We used the NIOX MINO (Aerocrine Inc, New Providence, NJ) to noninvasively measure FeNO. Based on previous research (ATS/ ERS, 2005), a questionnaire was administered to ascertain the following information prior to the FeNO measurement: (1) did the subject have a meal (breakfast if in the morning, lunch if in the afternoon) or exercise within an hour before the test; (2) was the subject's previous meal high in fat or sugar. An affirmative answer to any of these questions was tested as independent variables for their influence on FeNO and on air pollutant regression parameters. An scrubbing filter was used at the air intake to control for indoor nitric oxide.

2.4. Markers of systemic oxidative stress and inflammation

2.4.1. Oxidized low-density lipoprotein (oxLDL) and Interleukin-6 (IL-6)

At each follow-up visit, venous peripheral blood samples were drawn by a trained phlebotomist using anti-coagulant PPT BD Vacutainer[®] tubes and immediately centrifuged to separate plasma and transported at -20 °C before storage in our laboratory at -80 °C until analysis. The blood samples on one week (n=21) were lost due to improper storage procedures. The systemic oxidative stress biomarker oxLDL was measured in plasma by standardized ELISA using monoclonal antibodies directed against a neo-epitope in the aldehyde-substituted apoB-100 moiety of LDL (oxLDL-4E6 assay, Mercodia AB, Sweden). The lower limit of detection of oxLDL was 28.8 U/L and the results were considered invalid if the estimated CV was >25% (occurring in 0.34% of samples). Systemic inflammation biomarker IL-6 was measured in plasma using 96-well immunoassay kits (Quantikine High Sensitivity, Minneapolis, MN). The lower and upper limit of detection of IL-6 were 0.156 pg/ml and 10 pg/ml, respectively. IL-6 values were considered invalid if the estimated CV was > 25%. This occurred in 15.09% of samples.

2.5. Air pollution and meteorology

Targeted study areas included an approximate 18 km radius around the central air monitoring stations. Ambient hourly concentrations of PM_{2.5}, CO, nitrogen oxides (NO_x, nitric monoxide+ NO₂) and O₃ were ascertained from the South Coast Air Quality Management District (SCAQMD) monitoring stations (Appendix, Fig. A.2.). Measurements were based on Federal Reference Methods of the US EPA, including hourly PM_{2.5} from beta attenuation monitors (BAM). Hourly PM_{2.5} black carbon (BC, Aethelometer model AE22, Magee Scientific, Berkeley, CA) were measured at the University of Southern California (USC) monitoring sites near downtown Los Angeles and in Anaheim. Daily exposure data were calculated from the hourly data at the central air monitoring stations when \geq 75% of daily data were available. Missing rates for daily PM_{2.5}, CO, NO_x, O₃, and BC were 7.38%, 6.83%, 25.96%, 9.02%, and 3.28%, respectively. We used coefficients derived from regression modeling where measured exposures from the missing station were predicted by the exposure data from a neighboring station within the study area. All imputation linear models had $R^2 > 0.8$. If there were no other available data from nearby stations in the study area, then linear interpolation was used to impute missing data. We had a maximum of three continuous days of missing (except for BC in Los Angeles had one 8-day period that was missing). Less than 4% of the exposure data were linearly interpolated (Detailed information was presented in Appendix, Table A.1.). Ambient air pollutant concentrations for 1-day, 3-day, 5-day and 7-day averages preceding clinic follow-ups were calculated from the daily data. We preselected these averaging times to represent effect estimation for up to 7 days of exposure while avoiding an excessive presentation of data using all seven daily averaging times.

We also collected 5-day integrated concentrations of $PM_{0.18}$, $PM_{0.18-2.5}$ and $PM_{2.5-10}$ (MOUDI, model 100-1, MSP, Inc., Minneapolis, MN) at the USC monitoring sites near downtown Los Angeles and in Anaheim. The three particle size ranges represent

the ultrafine mode ($PM_{0.18}$), the accumulation mode ($PM_{0.18-2.5}$), and the coarse mode $(PM_{2.5-10})$. Although the typical size range for ultrafine particles have been stated in studies to be $\,< 0.1 \, \mu m$, the reported upper size cut point for particles that dominate the particle number concentration (which may be considered in the ultrafine mode) has varied temporally and spatially from 0.1 to $0.2 \,\mu m$ (Sioutas et al., 2005). We used five days of continuous particle collection because it was necessary to obtain a sufficient amount of PM mass loading for the chemical and oxidative potential assays described below and this duration had been previously demonstrated to be associated with cardiovascular and biomarker outcomes (such as IL-6) in our previous studies (Delfino et al., 2008, 2010a, 2010b). The USC monitoring site for Los Angeles was approximately 3 km southwest of the SCAQMD central air monitoring station where criteria air pollutants were measured, and was at the same location as the SCAQMD station in Anaheim (Appendix, Fig. A.2.). Carbonaceous species were measured in the ultrafine and accumulation mode fractions, and we included as exposure variables the sum of PAHs, the sum of hopanes (Appendix, Table A.2.), and elemental carbon (EC) as markers of various fossil fuel combustion products. Organic carbon (OC) was measured and is representative of both primary and secondary organic carbon sources. We also used the sum of organic (n-alkanoic) acids as a tracer of secondary organic aerosols (Rogge et al., 1993). For elemental species, we included the sum of available transition metals (V, Cr, Mn, Ni, Cu, and Fe) in all three sizefractions because they are known to participate in catalyzing Fenton's reaction that could generate oxidative stress (Ghio et al., 2012). Particle extracts were analyzed for oxidative potential [alveolar macrophage reactive oxygen species (ROS) assay and dithiothreitol (DTT) activity] as described below.

Following gravimetric measurements using a precision microbalance (Mettler Toledo Inc., Columbus, OH, USA) (± 0.001 mg), filters were sectioned and particles extracted for chemical characterization. EC and OC were quantified from a 1.5 cm² punch taken from the quartz/aluminum filter according to National Institute for Occupational Safety and Health Thermal Optical Transmission method (Schauer et al., 2003). PAHs and hopanes were analyzed using gas chromatography mass spectrometry (Stone et al., 2008). Total elemental composition of three size fractions was measured by digestion of a section of the Teflon filter-collected PM and subsequently analyzed by high-resolution sectorfield inductively-coupled plasma mass spectrometry (SF-ICPMS).

The Ogawa passive badge sampler (Ogawa & Co. USA, Inc. Pompano Beach, FL) was used to collect seven-day average personal exposures to NO_x . Subjects were instructed to wear the sampler clipped outside of clothing and placed near the bedside at night. Personal NO_x was collected on cellulose fiber filters and concentrations were determined by a spectrophotometer at a wavelength of 545 nm following the manufacturer's instructions (Ogawa & Company, 2006).

Meteorological data including temperature and relative humidity were obtained from central monitoring stations operated by SCAQMD. We calculated hourly heat index, the combined effects of temperature and relative humidity, using methods developed in a previous study (Anderson et al., 2013). Heat index was chosen because it is a better predictor of skin temperature, which is the main trigger of the body's cooling and warming mechanisms and thus is considered to be more closely related to health effects than temperature and/or relative humidity.

2.6. Measures of PM oxidative potential

Alveolar macrophage ROS assay represents the biotic oxidative potential of particle mixtures. Details of these methods are described elsewhere (Landreman et al., 2008). Briefly, biotic ROS production was quantified by extracting the 5-day composites of the three different size-fractionated PM filters with 1.00 ml of Milli-Q water. Rat alveolar macrophage cells (NR8383, American Type Culture Collection) in a 96-well plate were then exposed to PM extracts of both unfiltered (total ROS) and filtered (water-so-luble ROS, 0.22 µm polypropylene syringe filter). The fluorescent probe DCFH-DA (2',7'-dichloroluorescein diacetate) was used and the fluorescence intensity was measured using a plate reader, thus representing the cell-based oxidative generating capacity of PM. A model of microbial particles, un-opsonized Zymosan (a β -1,3-polysachharide of D-glucose) served as a positive ROS control as it binds to Toll-like receptor-2 on macrophage cells and then activates a strong respiratory burst and ROS production. ROS results are reported in µg Zymosan equivalent units per m³ air.

DTT activity represents the capacity of the PM extract to generate abiotic chemically-produced ROS (electron transfer from DTT to oxygen). DTT activity was quantified using a well-established method (Cho et al., 2005) on extracts of 5-day composites of the three different size-fractionated PM from quartz filters.

2.7. Statistical analysis

Data were analyzed using a linear mixed effect model to account for the clustering of longitudinal repeated measurements taken on each subject. All models include a random subject intercept to account for the correlation of repeated measures. We selected an autoregressive moving average (ARMA) (1,1) covariance structure based on Akaike's information criterion (AIC) for the covariance matrix to model the correlation between repeated measures in each subject. The magnitude of the effect was expressed relative to an interquartile range (25th to 75th percentile) difference in each pollutant concentration to compare associations across different pollutants.

2.7.1. Covariates

Time-invariant subject characteristics were controlled by the study design and the specified repeated measures model, and hence were not included as adjustment covariates. However, oxLDL levels were found to be correlated to LDL level. In order to test for the independent effect of oxLDL, we controlled for LDL level in the model of oxLDL. An a priori adjustment variable was heat index with the same lag average as the pollutant and was included in the model as a covariate to adjust for the potential confounding effect of weather. Time-dependent potential confounders were tested in the models, and were included in the model if they improved model fit. Days of gas stove use per week were significantly associated with increased personal NO_x and were adjusted in the models of personal NO_x a priori because the exposure of interest was outdoor fossil fuel sources of NO. To take into account seasonal variations in blood markers, long-term temporal trend (counts of the days during the study period) was tested by including cubic splines using differing numbers and locations of knots. However, this did not significantly change the parameter estimates of association or improve the model fit and so was not included in our final models. In exploratory models, exercise, food intake, sugar and fat intake prior to the clinic visit did not change the estimations or improve the model fit for FeNO and thus were not included in any models. All biomarker outcome variables were log-transformed before the analyses to fulfill the model assumption of residual normality. The results of biomarkers are given as the percent change in geometric mean.

2.7.2. Influential observations

The impact of influential observations was assessed using the Cook's D statistic and standardized residual diagnostics, at both the individual observations level and clustered subjects level (Cook, 1977). One influential subject (out of 88 subjects) was identified for EBC MDA and after excluding that subject the positive associations became stronger and more significant for BC, NO_x, CO, and ultrafine mass and components (Appendix, Table A.3). Three influential observations out of 719 observations for IL-6 were detected by restricted likelihood distance > 1 or Cook's D value > 4/n (n=number of observations) (Cook et al., 1982; Fox and Long, 1990). The estimated of regression coefficients for covariates of interest were generally unchanged but the confidence interval (CI) became smaller after excluding these three influential observations (Appendix, Table A.4).

2.7.3. Effect modification

In exploratory analyses we considered whether the following variables may be potential effect modifiers of the associations between biomarkers and air pollution: Age (\leq 75 years, >75 years), sex (male, female), obesity (body mass index: BMI \geq 30 kg/m²), history of hypertension (yes, no), intake of anti-hypertensive medication (yes, no), diabetes mellitus (yes, no), hypercholesterolemia by history (yes, no), total cholesterol/high-density lipoprotein (HDL) ratio (\geq 3.5, < 3.5) cardiovascular risk score (Goff et al., 2014) (\geq 0.2, < 0.2), intake of statin (yes, no), study area (Los Angeles, Anaheim), and former smokers (yes, no). Additionally, for airway biomarkers, we assessed effect modification by asthma and/or chronic obstructive pulmonary disease (COPD). Evidence of a significant interaction was considered at a nominal product term *p*-value < 0.1 to avoid increased type II errors in these hypothesis-generating analyses.

2.7.4. Sensitivity analysis

Several sensitivity analyses were conducted. First, to investigate potential exposure error, we restricted the analysis to the subjects who lived within the 90th percentile of subjects' residential distance to the stations (11.3 km for the SCAOMD monitoring stations and 13.1 km for the USC monitoring sites). Second, we limited the analysis to measured ambient exposures, rather than exposures including imputed data from the next nearest air monitoring station or interpolated from the adjacent day's data. Third, we excluded days with extreme heat index (< 52.5 and > 75.04, 10th percentile and 90th percentile of the heat index during the study period, respectively) because the decreased ventilation at subjects' residential buildings during those days may lead to increased exposure error for the representation of personal exposures by ambient measurements. Lastly, in order to test for the independent effects of primary and photochemically-produced air pollutants, we applied two-pollutant models by including a primary air pollutant (BC, NO_x and PAHs) and O₃, which showed unexpected inverse associations with some biomarkers. Note that this two-pollutant analysis is exploratory since O₃ and the primary pollutants were inversely correlated (Spearman $R \ge 0.6$). Interactions are also of interest because the highly oxidizing capability of O₃ give it the potential to react with primary pollutant components to produce other pro-oxidant components that can adversely affect the airways. In order to address this in an exploratory analysis we tested interactions between the same primary air pollutants and O₃ on the airway biomarkers (FeNO and EBC MDA).

All statistical analyses were performed using R version 3.0.3 (R Foundation for Statistical Computing, Vienna, Austria) or SAS 9.3 software (SAS, Cary, NC).

3. Results

3.1. Descriptive data

One hundred and ninety-one subjects (191) were recruited and

87 subjects dropped out or became ineligible (cancer, smoker, or moved out of the study area) before or soon after the start of follow-up. We additionally required a priori that subjects have at least 4 repeated outcome measures to provide sufficient withinsubject exposure-response data. We did a sensitivity analysis by including subjects with at least 2 repeated measurements and no fundamental changes in associations were observed. The median number of repeated measures for FeNO, EBC MDA, IL-6, and oxLDL are 11, 8, 8, and 10, respectively. Those subjects excluded from regression analyses due to lack of sufficient repeated outcomes were 16 subjects for EBC MDA, 8 subjects for FeNO, 11 subjects for IL-6, and 11 subjects for oxLDL. Among these listed above, 7 subjects were completely excluded for all outcomes leaving 97 with at least some outcome measurements. Among 97 subjects, 48 were from the Los Angeles area and 49 were from the Anaheim area. Descriptive characteristics of 97 subjects and descriptive data for the biomarker measurements are shown in Table 1 and the descriptive data by region are shown in Appendix Table A.5. The subjects were between 65 and 96 years old (mean, 74.8 years), with two-thirds of the subjects being female. Among the 97 subjects, there are 59 non-Hispanic Whites, 9 Hispanics, 11 African Americans, 9 Asians and 5 other race/ethnicities. Depending on the biomarker outcomes, the numbers of observations without any report of infection in the previous week ranged from 643 to 901 (Table 1). Spearman correlations showed no correlations (-0.04 to 0.11) among these 4 biomarkers. Additionally, we assessed the relationships among 4 biomarker outcomes in linear mixed models. No significant associations were observed (data not

Table 1

Characteristics of 97 subjects (Max n=946).

Characteristic	Mean \pm SD or N (%)
Age (years) \pm SD	74.8 ± 7.5
BMI $(kg/m^2) \pm SD$ Overweight (25–29.9) Obesity (\geq 30) Male Former smoker Cardiovascular disease Hypertension Hypercholesterolemia (by history)	$\begin{array}{c} 28.1 \pm 5.6 \\ 34 \ (35.1) \\ 31 \ (32.0) \\ 25 \ (25.8) \\ 42 \ (43.3) \\ 16 \ (16.5) \\ 66 \ (68.0) \\ 52 \ (53.6) \end{array}$
Lipid profile Total cholesterol > 200 mg/dL LDL > 130 mg/dL HDL Female < 50 mg/dL, male < 40 mg/dL Adult-onset diabetes mellitus COPD Asthma	$188.7 \pm 44.7 47 (48.5) 116.4 \pm 36.7 30 (30.9) 55.2 \pm 18.5 32 (33.0) 22 (22.7) 8 (8.3) 12 (12.4)$
Medications Anti-hypertensive medication HMG-CoA reductase inhibitors (statins)	63 (65.0) 45 (46.4)
Biomarkers IL-6 (pg/ml; n=716) OxLDL (U/L; n=841) FeNO (ppb; n=901) EBC MDA (nmol; n=643)	$\begin{array}{c} 2.4 \pm 1.7 \\ 59.2 \pm 20.2 \\ 26.6 \pm 13.4 \\ 8.4 \pm 5.1 \end{array}$

Abbreviations: BMI, body mass index; COPD: chronic obstructive pulmonary disease; EBC: exhaled breath condensate; FeNO: fractional concentration of exhaled nitric oxide; HDL: high-density lipoprotein; IL-6: Interleukin-6; LDL: high low-density lipoprotein; MDA: malondialdehyde; OxLDL: oxidized low-density lipoprotein.

shown).

Descriptive statistics of the exposures are shown in Table 2 and Appendix Table A.6 (by region). Data for specific transition metals are presented in the Appendix Table A.7. For 24-hour ambient pollutants, except for NO_x (missing rate=28%), all pollutants had less than 10% of data that were missing. The observations of personal NO had 13% missing due to subject noncompliance, incorrect sampling times, or measurement errors where NO concentration was less than NO₂. For size-fractionated PM components, up to 4 of the planned 48 weeks were missing due to equipment failure or power outages at the Anaheim site. Ninety-seven percent of the study period had average 24-hour PM_{2.5} below the National Air Ouality Standard recommended upper limit of $35 \,\mu\text{g/m}^3$. The concentration of total PAHs, hopanes, OC/EC, DTT and ROS were highest in PM_{0.18-2.5}. For total transition metals (the sum of V, Cr, Mn, Cu and Fe), the highest mass concentrations were observed in PM_{2.5-10}.

Correlations among traffic-related particulate air pollutants (BC, total PAHs in PM_{0.18} and PM_{0.18-2.5}, EC in PM_{0.18}) and between the traffic-related pollutants and total transition metals in PM_{0.18} and PM_{0.18-2.5} were strongly positive ($R \ge 0.64$, Table 3). These pollutants were also positively correlated with OC, which represents a mixture of primary and secondary organic aerosols. OC was positively correlated with organic acids (R=0.52 and 0.70 for PM_{0.18} and PM_{0.18-2.5}, respectively), which are tracers of secondary organic aerosols (Saffari et al., 2015). Traffic-related air pollutants were positively correlated with DTT, especially in PM_{0.18}, but not with ROS, and negatively correlated with O₃ and heat index (except EC_{0.18-2.5}).

3.2. Airway biomarkers and air pollutant associations

Associations between airway biomarkers and air pollutants are shown in Fig. 1. Since Daily ambient traffic-related pollutants (BC, CO, and NOx) were strongly correlated and showed similar results, we only display results of BC. PM within the ultrafine and accumulation mode ranges was of primary interest. Therefore, we only present PM components with these two size modes. For detailed information, see Appendix Table A.3 and Appendix Table A.8. The biomarker of airway oxidative stress, EBC MDA, was positively associated with daily ambient traffic-related air pollutants and 7-day personal NO_x, but not with BAM $PM_{2.5}$ (Fig. 1(A)). The strongest positive associations were observed for 7-day averages. For example, MDA had an estimated 7.4% increase (95% CI: 1.7%, 13.4%) per interquartile increase of 7-day average BC. For 5-day PM components, MDA was positively associated with total PM_{0.18} mass, carbonaceous aerosol components, and total transition metals (more strongly so with the PM_{0.18} fraction). We observed small but statistically nonsignificant positive associations of MDA with measurements of in vitro oxidative potential (ROS and DTT). The estimates of association for water-soluble ROS and total ROS were similar for all biomarkers, therefore, we do not present water-soluble ROS results. No associations were observed for organic acid while the associations of MDA with O₃ were negative.

The biomarker of airway inflammation, FeNO, was positively associated with markers of traffic-related air pollutants (BC, CO, NOx, and all carbonaceous aerosol components including OC), ultrafine $PM_{0.18}$ mass, and total transition metals (Fig. 1(B) and Appendix Table A.8.). For daily ambient measurements of traffic-related air pollutants (BC, CO, and NO_x), the strongest associations were observed for 7-day averages. For example, BC at 7-day average had the strongest estimated association (7.5%, 95% CI: 3.6%, 11.5%). For the 5-day average PM variables, we observed stronger estimated associations for ultrafine $PM_{0.18}$ than larger size-fractions for total mass, PAHs, hopanes, transition metals and organic acid. Hopanes in ultrafine $PM_{0.18}$ had the strongest

Table 2

Descriptive statistics of air pollutant measurements.

Pollutant	N (Missing)	Mean (SD)	IQR	Min	Max
Personal exposures (7-day average)					
NO _x (ppb)	820 (126)	29.3 (21.8)	21.3	2.2	160.1
Ambient exposures (24-hr Averages)					
Black Carbon (µg/m ³)	320 (16)	1.3 (0.9)	1.1	0.2	5.2
$PM_{2.5} (\mu g/m^3)$	309 (27)	17.6 (7.7)	9.6	3.8	49.2
CO (ppm)	311 (25)	0.5 (0.2)	0.3	0.1	1.6
Ozone (ppb)	303 (33)	23.2 (9.3)	12.7	1.3	48.5
NO _x (ppb)	241 (95)	35.4 (27.7)	31.0	3.6	175.6
Heat Index (F [°])	366 (0)	66.1 (8.4)	12.6	42.6	84.3
Size-fractionated PM (5-day average)					
Mass (µg/m ³)					
PM _{0.18}	45 (3)	2.4 (0.9)	1.1	1.2	4.8
PM _{0.18-2.5}	45 (3)	8.6 (3.2)	4.0	4.4	19.4
PM _{2.5-10}	44 (4)	14.9 (6.8)	7.8	4.5	35.4
Total PAHs (ng/m ³)					
PM _{0.18}	45 (3)	0.3 (0.2)	0.3	0.0	0.8
PM _{0.18-2.5}	45 (3)	0.5 (0.5)	0.5	0	2.0
Hopanes (ng/m ³)					
PM _{0.18}	45 (3)	0.2 (0.1)	0.2	0.0	0.4
PM _{0.18-2.5}	45 (3)	0.2 (0.2)	0.2	0	0.8
Organic acid $(\mu g/m^3)$					
PMo 18	45 (3)	26.2 (10.8)	14 5	96	58.0
PM _{0.18-2.5}	45 (3)	18.1 (14.0)	16.8	0.6	61.4
$OC(ug/m^3)$					
PMoto	45 (3)	12(04)	0.4	0.5	24
PMo.is	45 (3)	1.2(0.1) 15(0.8)	11	0.5	3.5
$PM_{2.5-10}$	45 (3)	0.7 (0.2)	0.3	0.3	1.1
$E((ug/m^3))$					
PMore	45 (3)	03(01)	0.2	01	0.6
PMoto of	45 (3)	0.3(0.1)	0.2	0.0	0.0
$PM_{2.5} = 10$	45 (3)	0.2(0.1)	0.1	0.0	0.09
1.002.5 - 10	13 (3)	0.0 (0.0)	0.1	0	0.05
Iotal ROS (µg Zym/m ³) ^a	45 (2)	10.0 (12.2)	10.0	21	52.2
PM _{0.18}	45 (3)	19.8 (12.2)	10.6	2.1	53.2
PM _{0.18} -2.5	45 (3)	136.7 (94.6)	125.9	26.2	394.0
PM _{2.5-10}	44 (4)	60.2 (49.7)	76.5	8.7	181.0
Dithiothreitol (nmol/min/m ³)					
PM _{0.18}	45 (3)	0.1 (0.1)	0.1	0.0	0.2
PM _{0.18-2.5}	45 (3)	0.3 (0.1)	0.1	0.1	0.5
$PM_{2.5-10}$	44 (4)	0.2 (0.1)	0.1	0.1	0.5
mansiion metais" (ng/m ²)	4E (2)	40.2 (40.4)	56.2	70	152 4
P1VI0.18	43 (<i>3</i>) 45 (2)	49.2 (40.4) 72.0 (52.0)	20.2	/.2	153.4
$F_{1VI_{0.18}-2.5}$	45 (3) 11 (1)	72.0 (52.0)	40.0 220.0	12.9	237.1
1 1¥12.5 – 10		330.0 (1/1.2)	220.0	112.0	514.0

Abbreviations: IQR: interquartile range; CO: carbon monoxide; PM: particulate matter; PAHs: polycyclic aromatic hydrocarbons; OC: organic carbon; EC: elemental carbon; ROS: Reactive oxygen species;

^a Zym: µg Zymosan equivalent units.

^b Total sum of transition metals include V, Cr, Mn, Ni, Cu and Fe.

association (9.4%, 95% CI: 5.2%, 13.6%). Associations between FeNO and DTT were positive, with borderline significance, while no association was observed for FeNO with ROS. Unexpectedly, FeNO was inversely associated with O_3 .

3.3. Systemic biomarkers and air pollutant associations

We did not observe significant associations between oxLDL and PM air pollutant components (Fig. 2(A) and Appendix Table A.9). However, positive (expected) but nonsignificant associations of oxLDL with traffic-related pollutants (BC, PAHs, and hopanes), total transition metals, ROS (especially in accumulation mode $PM_{0.18-2.5}$), and organic acids in $PM_{0.18-2.5}$, were observed. The systemic inflammation biomarker IL-6 was significantly and positively associated with daily ambient traffic-related air pollutants at 1-day averages and then gradually became weaker and

nonsignificant at 3-day, 5-day and 7-day averages (Fig. 2(B) and Appendix Table 4). We observed unexpected inverse associations with EC and with ROS in $PM_{0.18-2.5}$. The associations of IL-6 with 7-day personal NO_X and other 5-day average PM components were nonsignificant.

3.4. Effect modification and sensitivity analysis

We found evidence of potential effect modification by cardiovascular risk score on the associations between EBC MDA and air pollutants. Subjects with higher cardiovascular risk score (≥ 0.2) were more likely to have positive associations (Fig. 3). Consistent with this finding, the components of cardiovascular risk score: age (Appendix, Fig. A.3) and total cholesterol to HDL ratio (Appendix, Fig. A.4) also had similar modifying effects on the associations. The positive associations were pronounced in subjects who were older

Table 3 Spearman correlation matrix of selected pollutants³

	PM _{2.5}	BC	T PAHs _{0.18}	T PAHs 0.18-2.5	OC _{0.18}	OC _{0.18-2.5}	EC _{0.18}	EC _{0.18-2.5}	T ROS _{0.18}	T ROS _{0.18-2.5}	DTT 0.18	DTT 0.18-2.5	T metals _{0.18}	T metal _{0.18-2.5}	Heat index	Personal NO _x ^a
03	0.06	- 0.71	-0.84	- 0.81	- 0.57	- 0.71	- 0.60	-0.13	0.12	-0.07	-0.55	- 0.38	-0.61	-0.52	0.67	-0.43
PM _{2.5}		0.02	-0.12	-0.07	-0.17	0.15	0.08	0.19	0.05	0.45	-0.17	0.20	-0.27	-0.03	0.17	0.16
BC			0.92	0.92	0.84	0.89	0.82	0.47	0.18	0.24	0.79	0.62	0.81	0.72	-0.55	0.47
T PAHs _{0.18}				0.94	0.80	0.87	0.75	0.37	0.10	0.18	0.76	0.52	0.81	0.68	-0.70	I
T PAHS 0.18-2.5					0.73	0.85	0.64	0.32	- 0.02	0.09	0.68	0.49	0.76	0.63	-0.68	I
OC _{0.18}						0.76	0.83	0.46	0.45	0.26	0.86	0.61	0.80	0.70	-0.43	I
OC _{0.18-2.5}							0.76	0.47	0.12	0.31	0.63	0.62	0.64	0.63	-0.53	I
EC _{0.18}								0.58	0.42	0.39	0.77	0.57	0.68	0.66	-0.27	I
EC _{0.18-2.5}									0.34	0.41	0.50	0.25	0.32	0.27	0.09	I
T ROS _{0.18}										0.30	0.41	0.17	0.25	0.18	0.14	I
T ROS _{0.18-2.5}											0.14	0.37	0.13	0.30	0.12	I
DTT _{0.18}												0.44	0.81	0.68	-0.39	I
DTT _{0.18-2.5}													0.48	0.57	-0.39	I
T metals _{0.18} ^b														0.80	-0.54	I
T metals _{0.18-2.5}															-0.37	I
Heat index																-0.25
hhraviations: DTT- dit	hiothreit	ol. EC. al	mental carb	on. NO . nitroger	ovides.	OC oreani	c carbon	PM narti	culate mat	Per. BOS. read	tive ovvoe	cheriec. T.	total			

X. Zhang et al. / Environmental Research 150 (2016) 306-319

or had greater total cholesterol to HDL ratio. However, this effect modification by cardiovascular risk score was not observed in other outcome biomarkers.

We also found regional differences of the associations between IL-6 and 1-day average traffic-related air pollutants and 7-day average personal NO_X (Fig. 4(A)). Stronger positive associations were observed for traffic-related air pollutants in Los Angeles region than in Anaheim, while IL-6 was inversely associated with personal NO_x in Anaheim and positively but non-significantly associated with personal NO_X in Los Angeles. For O₃, significant inverse associations were observed in the Los Angeles, while positive but non-significant associations were observed in Anaheim. Regional differences were also observed between EBC MDA and daily traffic-related air pollutants and personal NOx. Positive associations were stronger in Anaheim than in Los Angeles (Fig. 4 (B)). However, these differences were not observed for the 5-day PM components. Other independent variables showed no statistically significant effect modification for any outcomes (data not shown).

Results from the sensitivity analyses did not qualitatively change the main findings when restricting subjects to a smaller study area radius around central air monitors, excluding extreme weather conditions, or excluding imputed ambient exposure data. Specifically, after excluding imputed BC exposure data, the estimated percent change of MDA per interquartile increase of 7-day average BC slightly decreased from 7.4% (95% CI: 1.7%, 13.4%) to 7.2% (1.1%, 13.4%). Results were similar for FeNO.

The estimates of association from two-pollutant models of O_3 with a primary air pollutant (BC, NOx and PAHs) became largely nonsignificant with effect estimates attenuating toward the null for airway biomarkers, except for associations between FeNO and PAHs in PM_{0.18} that remained significant. The results for the systemic biomarkers were similar using two-pollutant models and single-pollutant models (data not shown).

In the product term models testing interactions between ambient 1-day, 3-day, 5-day and 7-day averages of BC and O₃ on the airway biomarkers (FeNO and EBC MDA) we found significant positive interactions between the pollutants across all averaging times in relation to FeNO, with the strongest effect at the 5-day BC average (Fig. 5). Similar interactions on FeNO were also observed between O₃ and PAHs in both PM_{0.18} and PM_{0.18-2.5} (data not shown). However, this interaction effect was not observed for EBC MDA.

4. Discussion

were calculated with 7-day average of ambient pollutants and heat index

Total sum of transition metals include V, Cr, Mn, Ni, Cu and

Bold numbers indicate correlation values ≥ 0.60 and P < 0.05

Correlations for personal NO_x

4.1. Summary

We investigated the effects of short-term exposure to personal and ambient air pollutants, including their chemically-characterized components, on biomarkers of oxidative stress and inflammation in the airways and circulation in a susceptible elderly cohort in the Los Angeles air basin. We found evidence that pollutants with high oxidative potential, including markers of trafficrelated pollutants, ultrafine particles (PM_{0.18}), and transition metals, were associated with elevated airway oxidative stress (EBC MDA) and inflammation (FeNO). This is further evidenced by the positive, albeit non-significant, association between airway biomarkers and measurements of in vitro oxidative potential (ROS and DTT). In general, the associations for both of the airway biomarkers were stronger for traffic-related air pollutants in the ultrafine range than the larger particle size range. Moreover, we found acute positive associations between ambient traffic-related pollutants (1-day and 3-day pollutant averages of BC, NO_x and CO) and the biomarker of systemic inflammation (IL-6).



Fig. 1. Percent Change (mean and 95% confidence intervals) in airway inflammatory biomarkers EBC MDA (A) and airway oxidative stress biomarker FeNO (B) with a one interquartile range increase of ambient and personal air pollutants. Exposures were averaged across 5 days except as specified. BC: black carbon; CO: carbon monoxide; DTT: dithiothreitol; EBC MDA: malondialdehyde in exhaled breath condensate; EC: elemental carbon; FeNO: exhaled nitric oxide; NO_x: nitrogen oxides; OC: organic carbon; PAHs: polycyclic aromatic hydrocarbons; ROS: reactive oxygen species.

4.2. Airway biomarkers and air pollutant associations

MDA is a lipid peroxidation end-product and has been recognized as a biomarker of oxidative stress (Nordenhall et al., 2000). Higher levels of MDA in EBC have been observed in subjects with diseases characterized by oxidative stress, such as COPD (Lee et al., 2014) and diabetes (Dierckx et al., 2003). Few air pollution epidemiological studies have used EBC MDA as the biomarker of airway oxidative stress (Romieu et al., 2008; Gong et al., 2013; Sarnat et al., 2014). A study during the Beijing Olympic Games showed that a significant decrease in EBC MDA concentration in healthy young adults was associated with a substantial improvement in air quality (Gong et al., 2013). In a more recent study, EBC MDA was estimated to be higher three hours after the highway commutes of subjects during morning rush hours in the metropolitan Atlanta area in both asthmatic and non-asthmatic adults (Sarnat et al., 2014). However, to our knowledge, the present study is the first epidemiologic study that has shown positive associations of EBC MDA with components of aerosols with potentially high oxidative potential (ultrafine particle mass, PAHs, and transition metals) and with direct measures of PM oxidative potential (ROS and DTT).

Nitric oxide is produced in the respiratory tract as a part of the inflammatory process. It has previously been shown that Nitric oxide is increased in exhaled air with exposure to air pollution among patients with asthma (Delfino et al., 2006), COPD (Brindicci et al., 2005), other respiratory conditions (Mar et al., 2005), and coronary artery disease (Delfino et al., 2010a). Previous epidemiological studies have shown that FeNO is associated with

increases in traffic-related air pollutants in children (Delfino et al., 2006) and in young adults (Huang et al., 2012). However, this association has rarely been investigated in the general elderly population. We previously reported null associations of FeNO with traffic-related air pollutants (including markers of primary organic aerosols) in subjects with coronary artery disease, but found positive associations of FeNO with markers of secondary organic aerosols (Delfino et al., 2010a). In the present study, the estimated associations between FeNO and primary organic aerosols were positive, while we also found associations of FeNO with the one marker of secondary organic aerosols (organic acids) in PM_{0.18}, but not in PM_{0.18-2.5}. It is important to note that our previous study (Delfino et al., 2010a) had better exposure characterization including modeled hourly secondary organic carbon as a marker of secondary organic aerosol (Polidori et al., 2007). Another study of a group of elderly subjects in Steubenville, Ohio showed that FeNO was positively associated with exposure to 1-day PM_{2.5} mass concentrations (not chemically characterized). However, in our current study, we did not find a positive association between FeNO and PM_{2.5}. Instead, we found FeNO was positively associated with exposure to 5-day ultrafine PM_{0.18} mass, but not accumulation mode $PM_{0.18-2.5}$ mass concentrations. This may be due to the different chemical components in ultrafine versus accumulation mode size fractions of PM_{2.5} in Los Angeles as compared with Steubenville. Consistent with our current results, controlled human exposure studies have shown that acute exposure to diesel exhaust particles can transiently increase airway inflammation (Behndig et al., 2006; Bosson et al., 2008).



Fig. 2. Percent Change (mean and 95% confidence intervals) in systemic inflammatory biomarkers oxLDL (A) and systemic oxidative stress biomarker IL-6 (B) with a one interquartile range increase of ambient and personal air pollutants. Exposures were averaged across 5 days except as specified. BC: black carbon; CO: carbon monoxide; DTT: dithiothreitol; EC: elemental carbon; IL-6: Interleukin 6; NO_x: nitrogen oxides; OC: organic carbon; oxLDL oxidized low-density lipoprotein; PAHs: polycyclic aromatic hydrocarbons; ROS: reactive oxygen species.

4.3. Systemic biomarkers and air pollutant associations

Our findings of increased systemic inflammation with exposure to traffic-related air pollutants are generally consistent with previous studies (Rückerl et al., 2006, 2007; Delfino et al., 2009; Fang et al., 2012). However, the averaging time showing the strongest association has varied from hours to 9-days in these studies. It is unclear why different studies have shown different pollutant averaging times in association with biomarkers of systemic inflammation, but some combination of varying exposure error and population susceptibilities may be at play here.

Our previous study in elderly subjects with coronary artery disease reported significant associations of both airway and systemic inflammation with contrasting PM air pollutant characteristics (secondary organic aerosols and primary organic aerosols, respectively) (Delfino et al., 2010a). Notably, the air pollutant species that were significantly associated with airway versus systemic outcomes differed with the present study as discussed above. In contrast, our current study also revealed associations of biomarkers in the airways with markers of combustion-related air pollutants that were less clear for biomarkers of inflammation in the circulation. This may due to the fact that study subjects in the current cohort are relatively healthier than the subjects in the previous study. Healthy subjects may have stronger antioxidant defense mechanisms in their circulation that combats the effects of air pollution and chemical components that may spillover from the lungs. Furthermore, with considerably greater resources than the present study, the previous study measured air pollutants at the residential location (retirement communities) of each subject group that likely greatly reduced exposure error. Another possible explanation for why our current results for the systemic biomarkers are not as definitive as our previous studies could be due to the fact that the air pollutant levels, especially ultrafine PM, EC, OC, and tracers of mobile sources measured by our research group, have decreased over the last decade in the Los Angeles area (Shirmohammadi et al., 2016). This resulted from more stringent regulations on mobile source emissions by the United States EPA, California Air Resources Board and SCAQMD (Hasheminassab et al., 2014a).

4.4. Plausible biological mechanisms

In response to these pro-oxidant components in PM (trafficrelated pollutants, ultrafine PM, and transition metals), alveolar macrophages produce nitric oxide, which can combine with superoxide anion to produce peroxynitrite, a potent oxidizing compound, to generate ROS (Laumbach et al., 2010). Subsequently, excessive ROS can induce oxidative stress and inflammation in the airways. This is partly supported by our finding of positive, albeit nonsignificant, associations of both airway outcomes with *in vitro* macrophage ROS and DTT from ultrafine and accumulation mode PM extracts. This finding is consistent with our report of significant positive associations of FeNO with macrophage ROS and DTT from PM_{2.5} extracts in children and adolescents with asthma living in the Los Angeles air basin (Delfino et al., 2013).

The associations of PM oxidative potential (ROS and/or DTT)



Fig. 3. Effect modification of relations between exhaled breath condensate malondialdehyde (EBC MDA) and air pollution by cardiovascular score. Percent Change (mean and 95% confidence intervals) in EBC MDA with one interquartile range increase of selected air pollutant exposures averaged across 5 days (7 days for personal nitrogen oxides; NO_x) preceding each subject's clinic visits: *p < 0.1, **p < 0.0, compared with no effect modification by cardiovascular score.

with airway and systemic biomarkers were less strong than we expected. Nevertheless, the observed nonsignificant positive associations may be mediated by ROS-induced activation of *Nrf2* which is the major regulator of over 250 antioxidant and detoxifying enzymes and related proteins (Li et al., 2009). This potential mechanism is supported by our previous study that exposure markers of combustion-related air pollutants were positively associated with expression of the *Nrf2* gene (*NFE2L2*) and *Nrf2*-mediated genes (*HMOX1, NQO1,* and *SOD2*) in a cohort of elderly subjects with coronary artery disease (Wittkopp et al., 2015).

Previous epidemiological studies have shown that impaired vascular function is associated with short-term exposure to air pollutants, especially with pollutants linked to traffic sources (Ljungman et al., 2014; Provost et al., 2016). However, in our current study, we found relatively weak evidence of an elevation in systemic (plasma) oxidative stress and inflammation biomarkers with exposure to traffic-related pollutants. It is possible that these



Fig. 4. Effect modification of relations between interleukin-6 (IL-6) and air pollution (A), and exhaled breath condensate malondialdehyde (EBC MDA) and air pollution (B) by region. Percent Change (mean and 95% confidence intervals) in EBC MDA with one interquartile range increase of selected air pollutant exposures averaged across 1-day, 3-day, 5-day, and 7-day (7-day for personal nitrogen oxides; NO_x) preceding each subject's clinic visits: *p < 0.1, **p < 0.05, compared with no effect modification by region.

pollutants are translocated from the airways into the systemic circulation, where they may directly impact the endothelium and promote cardiovascular events without prior induction of systemic oxidative stress or inflammatory responses (Yamawaki et al., 2006; Wallenborn et al., 2007).

4.5. Effect modification

In assessing effect modification, we found suggestive evidence that subjects with a high cardiovascular risk score, as well as older subjects, and subjects with higher total cholesterol/HDL ratio experienced stronger and more significant positive associations of EBC MDA with air pollutants compared with healthier subjects. With the current study design, it is not possible to disentangle the independent effect of cardiovascular risk score, age or total cholesterol/HDL ratio. The stronger observed associations may simply be driven by single components of the cardiovascular risk score, such as age.

The observed regional differences for EBC MDA and IL-6 may be due to the fact that on average the baseline levels were different for the subjects in these two regions. For IL-6, the concentration in Los Angeles (1.95 pg/ml) is lower than in Anaheim (2.75 pg/ml) (Appendix Table A.5). While for EBC MDA, the average concentration in Los Angeles is 10.47 nmol, which was higher than in Anaheim (6.39 nmol). From the known medical conditions, we cannot explain different levels of these two biomarkers in Los Angeles and Anaheim, but the higher levels may be due to unmeasured factors that confound air pollution associations, which may explain the stronger associations of IL-6 and EBC MDA in



Ln(FeNO)= -0.0076*BC + -0.0063*O₃ + 0.0022*BC*O3 + -0.0008*heat index + 3.2979







Ln(FeNO)= -0.0414*BC + -0.0100*O₃ + 0.0054*BC*O3 + -0.0003*heat index + 3.3073 *p* value for interaction term = 0.0006





Fig. 5. Relation of FeNO (log value) to interaction between black carbon and O₃. Changes in log FeNO per unit increase of BC, O₃ and their interaction across 1 day (A), 3 days (B), 5 days (C), and 7 days (D) averaged before each subject's outcome measurement. BC: black carbon; FeNO: exhaled nitric oxide; O₃: ozone.

regions where their plasma concentrations were lower. It is important to note that in these exploratory analyses, we tested many interactions, hence multiple testing bias is a concern and the significant effect modification may be due to chance. However, we did not adjust for multiple comparisons as it is not explicitly required in exploratory analyses (Bender et al., 2001).

4.6. Pollutant interactions

The weakened associations for airway biomarkers with primary pollutants in the two-pollutant models with O_3 are likely due to the high inverse correlations between primary pollutants and O_3 (BC, NO_x and PAHs correlations with O_3 , $R \ge 0.6$). However, interactions may be of more interest because the highly oxidizing capability of O_3 give it the potential to react with primary pollutant components (such as those indicated by BC, NO_x and PAHs) leading to synergistic or additive adverse effects on human health. It was evidenced in an *in vitro* experimental study that PM and O_3 act synergistically in generating a sustained production of oxidative stress (Valavanidis et al., 2009). Consistent with these concepts we found positive interactions between primary pollutants and O_3 on FeNO. This may indicate that the underlying effect of air

pollutant mixtures may not be fully explained by only one air pollutant. Based on this observation, we built a model using ROS or DTT as the dependent variable and tested for interaction effects between BC/PAHs and O_3 . For BC and O_3 , the interaction effect was significant in Los Angeles but not in Anaheim. While for PAHs and O_3 , the interaction effect was significant in both areas, but stronger in Los Angeles than in Anaheim. This indicates that there may be other unmeasured secondary pollutant species that are formed during periods with both high primary air pollutants, and high O_3 that lead to higher oxidative potential of the particle mixtures.

4.7. Personal NO_x

Our results for personal NO_x were nonsignificant. The association between personal NO_x and EBC MDA was borderline significant but weaker than ambient 7-day NO_x (Appendix Table A.6). Meanwhile, we observed the strongest association of FeNO with ambient NO_x at 7-day average, but the relation was null with 7-day personal NOx (Appendix Table A.8). The lack of association for personal NO_x may be due to subject non-compliance, face velocity effects on the passive NO_x badge from variations in airflow, and unadjusted indoor sources of NO_x .

4.8. Strengths and limitations

The strengths of our study include the comprehensive characterization of PM components, including directly measuring PM oxidative potential using *in vitro* macrophage ROS and DTT, the repeated measurements of biomarkers over time, which minimize confounding from time-invariant between-subject and withinsubject factors, and the evaluation of biomarkers in both the airways and the circulation.

One of the limitations is that different biomarker outcomes were collected approximately at the same time. Therefore, we cannot investigate the relationships with respect to the time-sequence of oxidative stress and subsequent inflammatory responses to air pollution. Our size-fractionated PM composition data were collected across five-day integrated periods rather than daily limiting comparisons with ambient daily exposures that were measured continuously (BC and air pollutant gases). For example, we observed significant associations for IL-6 with continuously measured air pollutants at 1-day averages. Second, time-varying subject characteristics, such as daily physical activity and diet, can change associations between biomarkers and air pollution. However, we did not have this information in any detail resulting in potential unmeasured confounding. Third, measurement errors in exposure may have attenuated true associations. Most of our exposure data were obtained from central monitoring stations, which may not account for variations in personal exposure levels or variations in indoor air quality. Given that our elderly subjects spent most of their time indoors, our ambient outdoor exposure data may have introduced exposure error. Finally, external validity of the findings is limited to elderly subjects living in the selected regions of Los Angeles and Orange Counties.

5. Conclusions

Our results suggest that in a cohort of elderly adults, airway biomarkers of oxidative stress and inflammation are positively associated with traffic-related air pollutants, ultrafine particles and transition metals. We identified that pollutants with high oxidative potential, especially PM in the ultrafine range, have stronger estimated effects than in the larger PM size fractions. Acute and transient elevations of the systemic inflammatory biomarkers were observed with exposure to ambient daily trafficrelated pollutants. Our findings thus add mechanistic plausibility to the hypothesis that short-term exposure to traffic-related pollutants is associated with cardiorespiratory morbidity and mortality.

Conflict of interest

The authors declare that they have no conflict of interest.

Funding sources

This study was supported by grant number R01 ES12243 from the National Institute of Environmental Health Sciences, U.S. National Institutes of Health (NIH), and Grant UL1 TR000153 from the National Center for Research Resources and the National Center for Advancing Translational Sciences, NIH, and by contract numbers BPG-55175, BPG-53003, and AQMD-14172 administered by the South Coast Air Quality Management District (SCAQMD). The content is solely the responsibility of the authors and does not necessarily represent the official views of the NIH or SCAQMD.

Ethics approval

This study was approved by the institutional Review Board of the University of California, Irvine. All subjects provided written informed consent before participation.

Acknowledgements

We thank the staff and students at the University of California Irvine Department of Epidemiology, and staff at the Wisconsin State Laboratory of Hygiene for their assistance with the chemical analysis. We also acknowledge the support of University of Southern California's Provost and Viterbi PhD fellowships.

Appendix A. Supplementary material

Supplementary data associated with this article can be found in the online version at http://dx.doi.org/10.1016/j.envres.2016.06. 019.

References

- Anderson, G.B., Bell, M.L., Peng, R.D., 2013. Methods to calculate the heat index as an exposure metric in environmental health research. Environ. Health Perspect. 121 (10), 1111–1119. http://dx.doi.org/10.1289/ehp.1206273.
- ATS/ERS, 2005. ATS/ERS recommendations for standardized procedures for the online and offline measurement of exhaled lower respiratory nitric oxide and nasal nitric oxide, 2005. Am. J. Respir. Crit. Care Med. 171 (8), 912–930, DOI: 171/8/912 [pii] 10.1164/rccm.200406-710ST.
- Ayres, J.G., Borm, P., Cassee, F.R., Castranova, V., Donaldson, K., Ghio, A., Harrison, R. M., Hider, R., Kelly, F., Kooter, I.M., Marano, F., Maynard, R.L., Mudway, I., Nel, A., Sioutas, C., Smith, S., Baeza-Squiban, A., Cho, A., Duggan, S., Froines, J., 2008. Evaluating the toxicity of airborne particulate matter and nanoparticles by measuring oxidative stress potential - A workshop report and consensus statement. Inhal. Toxicol. 20 (1), 75–99. http://dx.doi.org/10.1080/ 08958370701665517.
- Behndig, A.F., Mudway, I.S., Brown, J.L., Stenfors, N., Helleday, R., Duggan, S.T., Wilson, S.J., Boman, C., Cassee, F.R., Frew, A.J., Kelly, F.J., Sandstrom, T., Blomberg, A., 2006. Airway antioxidant and inflammatory responses to diesel exhaust exposure in healthy humans. Eur. Respir. J. 27 (2), 359–365, DOI: 27/2/359 [pii] 10.1183/09031936.06.00136904.
- Bender, R., Lange, S., 2001. Adjusting for multiple testing when and how? J. Clin. Epidemiol. 54 (4), 343–349. http://dx.doi.org/10.1016/S0895-4356(00)00314-0.
- Borm, P.J.A., Kelly, F., Kunzli, N., Schins, R.P.F., Donaldson, K., 2007. Oxidant generation by particulate matter: from biologically effective dose to a promising, novel metric. Occup. Environ. Med. 64 (2), 73–74. http://dx.doi.org/10.1136/ oem.2006.029090.
- Bosson, J., Barath, S., Pourazar, J., Behindig, A.F., Sandstrom, T., Blomberg, A., Adelroth, E., 2008. Diesel exhaust exposure enhances the ozone-induced airway inflammation in healthy humans. Eur. Respir. J. 31 (6), 1234–1240. http://dx.doi. org/10.1183/09031936.00078407.
- Brindicci, C., Ito, K., Resta, O., Pride, N.B., Barnes, P.J., Kharitonov, S.A., 2005. Exhaled nitric oxide from lung periphery is increased in COPD. Eur. Respir. J. 26 (1), 52–59. http://dx.doi.org/10.1183/09031936.05.00125304.
- Brook, R.D., Rajagopalan, S., Pope 3rd, C.A., Brook, J.R., Bhatnagar, A., Diez-Roux, A. V., Holguin, F., Hong, Y., Luepker, R.V., Mittleman, M.A., Peters, A., Siscovick, D., Smith Jr., S.C., Whitsel, L., Kaufman, J.D., American Heart Association Council on Epidemiology and Prevention, Council on Kidney in Cardiovascular Disease, Council on Nutrition, Physical Activity and Metabolism, 2010. Particulate matter air pollution and cardiovascular disease: an update to the scientific statement from the American Heart Association. Circulation 121 (21), 2331–2378. http://dx.doi.org/10.1161/CIR.0b013e3181dbece1.
- Chen, R., Zhao, Z., Sun, Q., Lin, Z., Zhao, A., Wang, C., Xia, Y., Xu, X., Kan, H., 2015. Size-fractionated particulate air pollution and circulating biomarkers of inflammation, coagulation, and vasoconstriction in a panel of young adults. Epidemiology 26 (3), 328–336. http://dx.doi.org/10.1097/ EDE.00000000000273.
- Cho, A.K., Sioutas, C., Miguel, A.H., Kumagai, Y., Schmitz, D.A., Singh, M., Eiguren-Fernandez, A., Froines, J.R., 2005. Redox activity of airborne particulate matter at different sites in the Los Angeles Basin. Environ. Res. 99 (1), 40–47. http://dx. doi.org/10.1016/j.envres.2005.01.003.
- Cook, R.D., 1977. Detection of influential observation in linear-regression. Technometrics 19 (1), 15–18. http://dx.doi.org/10.2307/1268249.
- Cook, R.D., Weisberg, S., 1982. Residuals and Influence in Regression. Chapman & Hall, New York, NY.

- Daher, N., Hasheminassab, S., Shafer, M.M., Schauer, J.J., Sioutas, C., 2013. Seasonal and spatial variability in chemical composition and mass closure of ambient ultrafine particles in the megacity of Los Angeles. Environ. Sci.-Proc. Imp. 15 (1), 283–295. http://dx.doi.org/10.1039/c2em30615h.
- Delfino, R.J., Sioutas, C., Malik, S., 2005. Potential role of ultrafine particles in associations between airborne particle mass and cardiovascular health. Environ. Health Perspect. 113 (8), 934–946. http://dx.doi.org/10.1289/ehp.7938.
- Delfino, R.J., Staimer, N., Gillen, D., Tjoa, T., Sioutas, C., Fung, K., George, S.C., Kleinman, M.T., 2006. Personal and ambient air pollution is associated with increased exhaled nitric oxide in children with asthma. Environ. Health Perspect. 114 (11), 1736–1743.
- Delfino, R.J., Staimer, N., Tjoa, T., Polidori, A., Arhami, M., Gillen, D.L., Kleinman, M.T., Vaziri, N.D., Longhurst, J., Zaldivar, F., Sioutas, C., 2008. Circulating biomarkers of inflammation, antioxidant activity, and platelet activation are associated with primary combustion aerosols in subjects with coronary artery disease. Environ. Health Perspect, 116 (7), 898–906. http://dx.doi.org/10.1289/ehp.11189.
- Delfino, R.J., Staimer, N., Tjoa, T., Gillen, D.L., Polidori, A., Arhami, M., Kleinman, M.T., Vaziri, N.D., Longhurst, J., Sioutas, C., 2009. Air pollution exposures and circulating biomarkers of effect in a susceptible population: clues to potential causal component mixtures and mechanisms. Environ. Health Perspect. 117 (8), 1232–1238. http://dx.doi.org/10.1289/ehp.0800194.
- Delfino, R.J., Staimer, N., Tjoa, T., Arhami, M., Polidori, A., Gillen, D.L., George, S.C., Shafer, M.M., Schauer, J.J., Sioutas, C., 2010a. Associations of primary and secondary organic aerosols with airway and systemic inflammation in an elderly panel cohort. Epidemiology 21 (6), 892–902. http://dx.doi.org/10.1097/ EDE.0b013e3181f20e6c.
- Delfino, R.J., Staimer, N., Tjoa, T., Arhami, M., Polidori, A., Gillen, D.L., Kleinman, M.T., Schauer, J.J., Sioutas, C., 2010a. Association of biomarkers of systemic inflammation with organic components and source tracers in quasi-ultrafine particles. Environ. Health Perspect. 118 (6), 756–762. http://dx.doi.org/10.1289/ ehp.0901407.
- Delfino, R.J., Tjoa, T., Gillen, D.L., Staimer, N., Polidori, A., Arhami, M., Jamner, L., Sioutas, C., Longhurst, J., 2010b. Traffic-related air pollution and blood pressure in elderly subjects with coronary artery disease. Epidemiology 21 (3), 396–404. http://dx.doi.org/10.1097/EDE.0b013e3181d5e19b.
- Delfino, R.J., Staimer, N., Tjoa, T., Gillen, D.L., Schauer, J.J., Shafer, M.M., 2013. Airway inflammation and oxidative potential of air pollutant particles in a pediatric asthma panel. J. Expo. Sci. Environ. Epidemiol. 23 (5), 466–473. http://dx.doi. org/10.1038/jes.2013.25.
- Dierckx, N., Horvath, G., van Gils, C., Vertommen, J., van de Vliet, J., De Leeuw, I., Manuel-y-Keenoy, B., 2003. Oxidative stress status in patients with diabetes mellitus: relationship to diet. Eur. J. Clin. Nutr. 57 (8), 999–1008. http://dx.doi. org/10.1038/sj.ejcn.1601635.
- Elder, A., Oberdorster, G., 2006. Translocation and effects of ultrafine particles outside of the lung. Clin. Occup. Environ. Med. 5 (4), 785–796, DOI: S1526-0046 (06)00044-6 [pii] 10.1016/j.coem.2006.07.003.
- Fang, S.C., Mehta, A.J., Alexeeff, S.E., Gryparis, A., Coull, B., Vokonas, P., Christiani, D. C., Schwartz, J., 2012. Residential black carbon exposure and circulating markers of systemic inflammation in elderly males: the normative aging study. Environ. Health Perspect. 120 (5), 674–680. http://dx.doi.org/10.1289/ehp.1103982.
- Fox, J., Long, J.S., 1990. Modern Methods of Data-Analysis. pp. 257-291.
- Franck, U., Odeh, S., Wiedensohler, A., Wehner, B., Herbarth, O., 2011. The effect of particle size on cardiovascular disorders - The smaller the worse. Sci. Total Environ. 409 (20), 4217–4221. http://dx.doi.org/10.1016/j.scitotenv.2011.05.049.
- Franklin, B.A., Brook, R., Pope 3rd, C. Arden, 2015. Air pollution and cardiovascular disease. Curr. Probl. Cardiol. 40 (5), 207–238. http://dx.doi.org/10.1016/j. cpcardiol.2015.01.003.
- Ghio, A.J., Carraway, M.S., Madden, M.C., 2012. Composition of air pollution particles and oxidative stress cells, tissues, living system. J. Toxicol. Environ. Heal B 15 (1), 1–21. http://dx.doi.org/10.1080/10937404.2012.632359.
- Goff, D.C., Lloyd-Jones, D.M., Bennett, G., Coady, S., D'Agostino, R.B., Gibbons, R., Greenland, P., Lackland, D.T., Levy, D., O'Donnell, C.J., Robinson, J.G., Schwartz, J. S., Shero, S.T., Smith, S.C., Sorlie, P., Stone, N.J., Wilson, P.W.F., 2014. 2013 ACC/ AHA guideline on the assessment of cardiovascular risk a report of the American College of Cardiology/American Heart Association Task Force on practice guidelines. J. Am. Coll. Cardiol. 63 (25), 2935–2959. http://dx.doi.org/10.1016/j. jacc.2013.11.005.
- Gong, J.C., Zhu, T., Kipen, H., Wang, G.F., Hu, M., Ohman-Strickland, P., Lu, S.E., Zhang, L., Wang, Y.D., Zhu, P., Rich, D.Q., Diehl, S.R., Huang, W., Zhang, J.F., 2013. Malondialdehyde in exhaled breath condensate and urine as a biomarker of air pollution induced oxidative stress. J. Expo. Sci. Environ. Epidemiol. 23 (3), 322–327. http://dx.doi.org/10.1038/jes.2012.127.
- Hasheminassab, S., Daher, N., Ostro, B.D., Sioutas, C., 2014a. Long-term source apportionment of ambient fine particulate matter (PM2.5) in the Los Angeles Basin: a focus on emissions reduction from vehicular sources. Environ. Pollut. 193, 54–64. http://dx.doi.org/10.1016/j.envpol.2014.06.012.
- Hasheminassab, S., Daher, N., Shafer, M.M., Schauer, J.J., Delfino, R.J., Sioutas, C., 2014b. Chemical characterization and source apportionment of indoor and outdoor fine particulatematter (PM2.5) in retirement communities of the Los Angeles Basin. Sci. Total Environ. 490, 528–537. http://dx.doi.org/10.1016/j. scitotenv.2014.05.044.
- Holvoet, P., Harris, T.B., Tracy, R.P., Verhamme, P., Newman, A.B., Rubin, S.M., Simonsick, E.M., Colbert, L.H., Kritchevsky, S.B., 2003. Association of high coronary heart disease risk status with circulating oxidized LDL in the wellfunctioning elderly: findings from the health, aging, and body composition study. Arter. Thromb. Vasc. Biol. 23 (8), 1444–1448, DOI: 10.1161/01.

ATV.0000080379.05071.22 01.ATV.0000080379.05071.22 (pii).

- Huang, W., Wang, G., Lu, S.E., Kipen, H., Wang, Y., Hu, M., Lin, W., Rich, D., Ohman-Strickland, P., Diehl, S.R., Zhu, P., Tong, J., Gong, J., Zhu, T., Zhang, J., 2012. Inflammatory and oxidative stress responses of healthy young adults to changes in air quality during the Beijing Olympics. Am. J. Crit. Care 186 (11), 1150–1159. http://dx.doi.org/10.1164/rccm.201205-08500C.
- Landreman, A.P., Shafer, M.M., Hemming, J.C., Hannigan, M.P., Schauer, J.J., 2008. A macrophage-based method for the assessment of the reactive oxygen species (ROS) activity of atmospheric particulate matter (PM) and application to routine (daily-24 h) aerosol monitoring studies. Aerosol Sci. Technol. 42 (11), 946–957. http://dx.doi.org/10.1080/02786820802363819.
- Lanzinger, S., Schneider, A., Breitner, S., Stafoggia, M., Erzen, I., Dostal, M., Pastorkova, A., Bastian, S., Cyrys, J., Zscheppang, A., Kolodnitska, T., Peters, A., Grp, U.S., 2016. Associations between ultrafine and fine particles and mortality in five central European cities - results from the UFIREG study. Environ. Int. 88, 44–52. http://dx.doi.org/10.1016/j.envint.2015.12.006.
- Larstad, M., Ljungkvist, G., Olin, A.C., Toren, K., 2002. Determination of malondialdehyde in breath condensate by high-performance liquid chromatography with fluorescence detection. J. Chromatogr. B 766 (1), 107–114. http: //dx.doi.org/10.1016/S0378-4347(01)00437-6.
- Laumbach, R.J., Kipen, H.M., 2010. Acute effects of motor vehicle traffic-related air pollution exposures on measures of oxidative stress in human airways. Ann. NY Acad. Sci. 1203, 107–112. http://dx.doi.org/10.1111/j.1749–6632.2010.05604.x.
- Lee, J.S., Shin, J.H., Hwang, J.H., Baek, J.E., Choi, B.S., 2014. Malondialdehyde and 3-nitrotyrosine in exhaled breath condensate in retired elderly coal miners with chronic obstructive pulmonary disease. Saf. Health Work. 5 (2), 91–96. http://dx.doi.org/10.1016/j.shaw.2014.03.001, S2093-7911(14)00019-5 (pii).
- Li, N., Kim, S., Wang, M., Froines, J., Sioutas, C., Nel, A., 2002. Use of a stratified oxidative stress model to study the biological effects of ambient concentrated and diesel exhaust particulate matter. Inhal. Toxicol. 14 (5), 459–486. http://dx. doi.org/10.1080/089583701753678571.
- Li, N., Sioutas, C., Cho, A., Schmitz, D., Misra, C., Sempf, J., Wang, M.Y., Oberley, T., Froines, J., Nel, A., 2003. Ultrafine particulate pollutants induce oxidative stress and mitochondrial damage. Environ. Health Perspect. 111 (4), 455–460. http: //dx.doi.org/10.1289/ehp.6000.
- Li, W.G., Kong, A.N., 2009. Molecular mechanisms of Nrf2-mediated antioxidant response. Mol. Carcinog. 48 (2), 91–104. http://dx.doi.org/10.1002/mc.20465.
- Lippmann, M., 1977. Regional deposition of particles in the human respiratory trend. In: Lee, D.H.K., Falk, H.L., et al., Handbook of physiology, reaction to environmental agents. American Physiological Society, Bethesda, MD.
- Ljungman, P.L., Wilker, E.H., Rice, M.B., Schwartz, J., Gold, D.R., Koutrakis, P., Vita, J. A., Mitchell, G.F., Vasan, R.S., Benjamin, E.J., Mittleman, M.A., Hamburg, N.M., 2014. Short-term exposure to air pollution and digital vascular function. Am. J. Epidemiol. 180 (5), 482–489. http://dx.doi.org/10.1093/aje/kwu161.
- Mar, T.F., Jansen, K., Shepherd, K., Lumley, T., Larson, T.V., Koenig, J.Q., 2005. Exhaled nitric oxide in children with asthma and short-term PM2.5 exposure in Seattle. Environ. Health Perspect. 113 (12), 1791–1794.
- Nordenhall, C., Pourazar, J., Blomberg, A., Levin, J.O., Sandstrom, T., Adelroth, E., 2000. Airway inflammation following exposure to diesel exhaust: a study of time kinetics using induced sputum. Eur. Respir. J. 15 (6), 1046–1051.
- Ogawa & Company, U., Inc. 2006. NO, NO2, NOx and SO2 Sampling Protocol Using the Ogawa Sampler.
- Pant, P., Harrison, R.M., 2013. Estimation of the contribution of road traffic emissions to particulate matter concentrations from field measurements: a review. Atmos. Environ. 77, 78–97. http://dx.doi.org/10.1016/j.atmosenv.2013.04.028.
- Polidori, A., Arhami, M., Sioutas, C., Delfino, R.J., Allen, R., 2007. Indoor/outdoor relationships, trends, and carbonaceous content of fine particulate matter in retirement homes of the Los Angeles basin. J. Air Waste Manag. 57 (3), 366–379.
- Provost, E.B., Louwies, T., Cox, B., Op't Roodt, J., Solmi, F., Dons, E., Panis, L. Int, Boever, P. De, Nawrot, T.S., 2016. Short-term fluctuations in personal black carbon exposure are associated with rapid changes in carotid arterial stiffening. Environ. Int. 88, 228–234. http://dx.doi.org/10.1016/j.envint.2015.12.023.
- Rogge, W.F., Mazurek, M.A., Hildemann, L.M., Cass, G.R., Simoneit, B.R.T., 1993. Quantification of urban organic aerosols at a molecular-level - identification, abundance and seasonal-variation. Atmos. Environ. A-Gen. Topics 27 (8), 1309–1330. http://dx.doi.org/10.1016/0960–1686(93)90257-Y.
- Romieu, I., Barraza-Villarreal, A., Escamilla-Nunez, C., Almstrand, A.C., Diaz-Sanchez, D., Sly, P.D., Olin, A.C., 2008. Exhaled breath malondialdehyde as a marker of effect of exposure to air pollution in children with asthma. J. Allergy Clin. Immun. 121 (4), 903–909. http://dx.doi.org/10.1016/j.jaci.2007.12.004.
- Ruckerl, R., Hampel, R., Breitner, S., Cyrys, J., Kraus, U., Carter, J., Dailey, L., Devlin, R. B., Diaz-Sanchez, D., Koenig, W., Phipps, R., Silbajoris, R., Soentgen, J., Soukup, J., Peters, A., Schneider, A., 2014. Associations between ambient air pollution and blood markers of inflammation and coagulation/fibrinolysis in susceptible populations. Environ. Int. 70, 32–49. http://dx.doi.org/10.1016/j.envint.2014.05.013.
- Civini, R., Ibald-Mulli, A., Koenig, W., Schneider, A., Woelke, G., Cyrys, J., Heinrich, J., Marder, V., Frampton, M., Wichmann, H.E., Peters, A., 2006. Air pollution and markers of inflammation and coagulation in patients with coronary heart disease. Am. J. Respir. Crit. Care Med. 173 (4), 432–441, DOI: 200507-11230C [pii] 10.1164/rccm.200507-11230C.
- Rückerl, R., Greven, S., Ljungman, P., Aalto, P., Antoniades, C., Bellander, T., Berglind, N., Chrysohoou, C., Forastiere, F., Jacquemin, B., von Klot, S., Koenig, W., Kuchenhoff, H., Lanki, T., Pekkanen, J., Perucci, C.A., Schneider, A., Sunyer, J., Peters, A., 2007. Air pollution and inflammation (interleukin-6,C-reactive protein, fibrinogen) in myocardial inflarction survivors. Environ. Health Perspect. 115 (7),

1072-1080. http://dx.doi.org/10.1289/ehp.10021.

- Saffari, A., Hasheminassab, S., Wang, D.B., Shafer, M.M., Schauer, J.J., Sioutas, C., 2015. Impact of primary and secondary organic sources on the oxidative potential of quasi-ultrafine particles (PM0.25) at three contrasting locations in the Los Angeles Basin. Atmos. Environ. 120, 286–296. http://dx.doi.org/10.1016/j. atmosenv.2015.09.022.
- Sarnat, J.A., Golan, R., Greenwald, R., Raysoni, A.U., Kewada, P., Winquist, A., Sarnat, S.E., Flanders, W.D., Mirabelli, M.C., Zora, J.E., Bergin, M.H., Yip, F., 2014. Exposure to traffic pollution, acute inflammation and autonomic response in a panel of car commuters. Environ. Res. 133, 66–76. http://dx.doi.org/10.1016/j. envres.2014.05.004.
- Schauer, J.J., Mader, B.T., Deminter, J.T., Heidemann, G., Bae, M.S., Seinfeld, J.H., Flagan, R.C., Cary, R.A., Smith, D., Huebert, B.J., Bertram, T., Howell, S., Kline, J.T., Quinn, P., Bates, T., Turpin, B., Lim, H.J., Yu, J.Z., Yang, H., Keywood, M.D., 2003. ACE-Asia intercomparison of a thermal-optical method for the determination of particle-phase organic and elemental carbon. Environ. Sci. Technol. 37 (5), 993–1001. http://dx.doi.org/10.1021/es020622f.
- Shirmohammadi, F., Hasheminassab, S., Saffari, A., Schauer, J.J., Delfino, R.J., Sioutas, C., 2016. Fine and ultrafine particulate organic carbon in the Los Angeles basin: trends in sources and composition. Sci. Total Environ. 541, 1083–1096. http: //dx.doi.org/10.1016/j.scitotenv.2015.09.133.
- Sioutas, C., Delfino, R.J., Singh, M., 2005. Exposure assessment for atmospheric ultrafine particles (UFPs) and implications in epidemiologic research. Environ. Health Perspect. 113 (8), 947–955.
- Stone, E.A., Snyder, D.C., Sheesley, R.J., Sullivan, A.P., Weber, R.J., Schauer, J.J., 2008. Source apportionment of fine organic aerosol in Mexico City during the MI-LAGRO experiment 2006. Atmos. Chem. Phys. 8 (5), 1249–1259.

- Valavanidis, A., Loridas, S., Vlahogianni, T., Fiotakis, K., 2009. Influence of ozone on traffic-related particulate matter on the generation of hydroxyl radicals through a heterogeneous synergistic effect. J. Hazard Mater. 162 (2–3), 886–892. http: //dx.doi.org/10.1016/i.jhazmat.2008.05.124.
- Wallenborn, J.G., McGee, J.K., Schladweiler, M.C., Ledbetter, A.D., Kodavanti, U.P., 2007. Systemic translocation of particulate matter-associated metals following a single intratracheal instillation in rats. Toxicol. Sci. 98 (1), 231–239. http://dx. doi.org/10.1093/toxsci/kfm088.
- Weichenthal, S., 2012. Selected physiological effects of ultrafine particles in acute cardiovascular morbidity. Environ. Res. 115, 26–36. http://dx.doi.org/10.1016/j. envres.2012.03.001.
- Wittkopp, S., Staimer, N., Tjoa, T., Stinchcombe, T., Daher, N., Schauer, J.J., Shafer, M. M., Sioutas, C., Gillen, D.L., Delfno, R.J., 2015. Nrf2-related gene expression and exposure to traffic-related air pollution in elderly subjects with cardiovascular disease: an exploratory panel study. J. Expo. Sci. Environ. Epidemiol. http://dx. doi.org/10.1038/jes.2014.84
- Wu, S., Yang, D., Wei, H., Wang, B., Huang, J., Li, H., Shima, M., Deng, F., Guo, X., 2015. Association of chemical constituents and pollution sources of ambient fine particulate air pollution and biomarkers of oxidative stress associated with atherosclerosis: a panel study among young adults in Beijing, China. Chemosphere 135, 347–353, DOI: 10.1016/j.chemosphere.2015.04.096 S0045-6535(15) 00454-3 (pii).
- Yamawaki, H., Iwai, N., 2006. Mechanisms underlying nano-sized air-pollutionmediated progression of atherosclerosis: carbon black causes cytotoxic injury/ inflammation and inhibits cell growth in vascular endothelial cells. Circ. J. 70 (1), 129–140.