We would like to thank the referees for all the comments. The authors' responses are listed as
 follows with corresponding changes made in the manuscript.

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5 Anonymous Referee #1

6 In this study, the authors assessed the water-soluble oxidative potential of PM2.5 collected 7 in Southeastern U.S. based on both DTT and AA assays, and compared the results of two assays 8 in the view of their association with chemical components, sources and emergency department 9 (ED) visits. In my opinion, this is an important and careful study with large database, providing 10 essential information on the origin and potential health outcome of the water-soluble oxidative 11 potentials of PM2.5. In addition, the result could help future studies to better interpret the data 12 13 based on those assays. However, there are several issues that should be addressed in the manuscript. A major concern is on the oxidative potential of PM2.5. Actually I'm afraid that 14

15 oxidative potential from water-soluble components is far from enough for the evaluation of

- 16 PM2.5's toxicity or health effects, given that a series of studies have suggested that some
- hydrophobic components (e.g. represented by PAHs) from vehicles are the major toxic
 components on human health (Delfino et al., Environ Health Perspect 2010, 118: 756-762; etc.).

Why not measure the oxidative potential of hydrophobic components if the authors wanted to

20 link the oxidative potential of PM2.5 with some health outcomes?

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22 Authors' response: This paper does focus only on the water-soluble PM components

contributing to oxidative potential, and doesn't include the water-insoluble fraction. There is no doubt that water-insoluble components also have potential adverse health impacts. There is a logical reason to look separately at soluble vs insoluble components, however, since they are likely to have differing physiological effects once deposited onto the aqueous environment of the air-lung interface. Although we decided to focus solely on soluble species, we are not implying that insoluble species are un-important, it is just not the focus of this work. Besides, there is no standard protocol so far for measuring the water-insoluble ROS activities.

Changes in manuscript: Added to the Introduction section "Although in-soluble components
are important, since there is no current standard protocols for measuring the water-insoluble
oxidative potential, we focus solely on the water-soluble AA and DTT activities."

- 35 The following are some specific comments:
 - 1. Page 30615, section 2.2.1: Could sonication for half an hour in the water phase generate OH radical, which could result in great oxidative potential?

Authors' response: There is some concern on the alteration of the sample due to OH
formation during sonication [*Miljevic et al.*, 2014]. We have conducted a limited number of
experiment on the effect of sonication on the measured oxidative potential by comparing
water-soluble DTT activity from the same extracts after sonication vs shaking for 3hrs. Our
findings suggest little difference between the two. Therefore, although sonication may
generate OH radicals, it doesn't seem to have significant effects on our DTT assay results. A
manuscript on the results is in preparation.

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47 48 49 50 51 52	Changes in manuscript: page 30617 line 14, added "Although OH may form during sonication (Miljevic et al., 2014), it has little effect on our ROS measurement since we compared the water-soluble ROS activities from the same sample that had been extracted by shaking for 3 hours vs sonication and found no significant differences, average ratio is 1.08 ± 0.20 , n = 7."	
53 54 55 56	 Page 30625, line 15: The spatio-temporal analysis could not draw the conclusion that the oxidative potential is influenced by different components from different sources, because there are no evidences in this part showed the similar trend for chemical components and sources. 	
57 58	Authors' response: We have removed the reference to chemical components.	
59 60 61 62 63 64	Changes in manuscript: Page 30625 line 15: changed "These results indicate that water- soluble DTT and AA activities of PM2.5 were influenced by different aerosol components from differing sources that varied with season." to "These results indicate that there are differences in the sources for water-soluble AAv and DTTv, with traffic emissions apparently a more significant source for AAv. Correlation analysis with specific aerosol components provides further insights."	
65 66 67 68	3. Page 30626, line1: r2 or r? In case of r2, it's better to convert it into r, since r is used throughout the manuscript.	
69 70	Changes in manuscript: page 30626 line 1: changed "r2>0.5" to "r>0.7".	
71 72 73	4. Page 30626, line 25: Since Pearson's r is used. Please provide information on the normality of the data and on whether data was log-transformed.	
74 75 76 77 78 79 80	Authors' response: As is found for almost all atmospheric constituents, their frequency distributions are log-normally distributed. However, the correlation analyses were performed using measured concentrations rather than log-transformed concentrations as this is a straightforward interpretation method for these data. Since interpretation of the correlation analysis results is similar whether done on a linear or log basis, we prefer to keep the former here.	
81 82 83	Changes in manuscript: No changes.	
84 85 86	 Page 30630, section 3.2.2: It is better to provide detailed data of the risk ratio (together with 95% CI) in the text instead of in Fig. 4 only. 	
87 88 89 90	Authors' response: The risk ratios for AA for asthma/wheeze and congestive heart failure are added to the text. The detailed data for other health outcomes are provided as supplemental material.	
91 92	Changes in manuscript: Page 30630, line 13, changed "For asthma/wheeze and congestive heart failure, the risk ratios for an increase of an interquartile range for DTT estimated from	

93 the model (Eq. 6) were above 1 (Fig. 4), and the 95% confidence intervals exclude 1, 94 indicating a statistically significant positive association between estimated DTT and the ED visits for these health outcomes." to "For asthma/wheeze and congestive heart failure, 95 96 although the risk ratios for an increase of an interquartile range for AA_v^{P} were above 1 [1.005 and 1.003 for Asthma/wheeze and CHF, respectively, Fig. 5(a)], the 95 % confidence 97 intervals crossed 1 (0.994-1.015 and 0.986-1.020 for Asthma/wheeze and congestive heart 98 99 failure, respectively), indicating a non-statistically significant association between AA_v^{e} and 100 the ED visits for these health outcomes." 101

6. Page 30631, line 3: The authors claimed that the epidemiology analysis "support aerosol 104 particle oxidative potential as a mechanism contributing to these PM-induced adverse 105 106 health effects", which I think might not be true. A more robust association doesn't necessarily mean a possible mechanism. For example, sulfate is considered as "benign" 107 (page 30612, line 13), but actually sulfate is strongly associated with adverse health 108 effects in epidemiology studies, probably because sulfate was co-emitted with toxic 109 pollutants (Grahame, EHP, 2012). Personally, I think the association of DTT could also 110 111 possibly be attributed to co-emission with toxic pollutants, especially given that (1). DTT is sensitive to organic species; (2). There is a lack of toxicology studies showing the 112 health effects of oxidative potential; (3). No significant association was observed for AA. 113 Although the authors attributed it to the different uncertainties (page 30630, line 19), it is 114 not convincing since r values of 0.60 and 0.68 are not that different. 115

Authors' response:

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First, we state that it "supports", not that it proves OP is a mechanism. We would like to address the reviewer's three comments one by one:

1) DTT is sensitive to organic species

DTT assay is sensitive to organic species doesn't prove that OP is not a plausible mechanistic linkage. We've identified components of the organic species that are DTT active [*Verma et al.*, 2015a; *Verma et al.*, 2015b]. For example, HULIS (humic-like substances) in biomass burning or more-oxidized OA are mostly linked to DTT while biogenic secondary organics basically have no DTT activity.

- 2) There is a lack of toxicology studies showing the health effects of oxidative potential
 This is not true. There are numbers of studies have associated OP to PM toxicity, as we
 have cited in the introduction section (see page 30612 line 27).
- 3) No significant association was observed for AA. Although the authors attributed it to the
 different uncertainties (page 30630, line 19), it is not convincing since r values of 0.60
 and 0.68 are not that different
- We have rearranged the text to clarify the differences in health analysis between DTT andAA are possibly due to the narrower selectivity of AA than DTT assay, rather than the
- uncertainties in the models, which is fairly similar (ie, r values 0.60 vs 0.68).

139 140 In summary, although a positive association does not mean a causation, the association of 141 DTT with health effects is supported by a rational physiological mechanism (the DTT assay 142 measures the OP of particles, which represents the ability of PM components to generate ROS and ROS can induce oxidative stress in vivo. It is also supported by our various findings 143 that DTT is sensitive to specific classes of organic species and metals that could lead to 144 145 health effects. 146 **Changes in manuscript:** 147 148

Page 30630 line 18: changed "Similar differences were found for estimates based..." to "The 149 same results were found for estimates based on the two other regressions [(see Figure 5(b) 150 and (c)]), suggesting that the null relationship of AA_v^e and positive association of DTT_v^e with 151 these health outcomes are to some extent robust, despite the high uncertainties from the back-152 cast models. A possible cause for the differences in AA_v^e and DTT_v^e health associations is the 153 more narrow selectivity of the AA assay to specific aerosol components (i.e., mostly 154 sensitive to Cu). The AA assay may not capture the overall oxidative potential of all the 155 various PM components as well as the DTT assay." 156 157

- 159 Referee # 2 Cort Anastasio
- 161 Introduction.

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162 Fang and co-authors describe results from two different measurements of oxidative potential -

163 DTT (dithiothreitol) and AA (ascorbic acid) – for ambient particles collected in the southeast

U.S. as part of the SCAPE center. The DTT and AA data sets are analyzed in three different ways: (1) linear regressions between oxidative potential and various chemicals in an attempt to

identify the responsible chemical species; (2) positive matrix factorization (PMF) and chemical

mass balance (CMB) modeling to identify important sources of ROS-generating PM; and (3)

epidemiological modeling on approximately a decadelong time series of estimated DTT and AA

results to assess if either measure of oxidative potential is associated with health effects.

170 Authors' response: We also investigated seasonal trends and spatial distributions.

171 The AA results are novel and, though not fully explored, form the core of a good manuscript.

172 With some additional pieces, this could be a very nice piece of work. On the other hand, the DTT

173 results have all been presented previously and there is nothing that warrants spending half the

174 manuscript on these past results. There are some new (and better) ways in which the DTT data

could be treated; if this is done, it could be an important contribution to our understanding ofDTT and would significantly improve the manuscript.

- 177 Authors' response: First, we stress that this is a comparison paper. Our aim is to provide a
- clear contrast between the AA vs DTT assays, rather than simply discuss AA results and
- 179 leaving it the reader to go back and forth between multiple papers. Admittedly, the AA
- 180 results do not show strong linkages to health (ie, a null result), but this is an important result.
 181 However, the real strength of this paper is to provide a direct side by side comparison.
- 181 However, the real strength of this paper is to provide a direct side-by-side comparison

between AA and DTT, which were both analyzed on the same filters. That is, we go through 182 a whole series of comparisons: spatial, seasonal, correlations to PM components, including 183 PM_{2.5} mass (important since it has been associated with adverse health effects in many 184 studies and is regulated), source apportionment, and then health effects. We have modified 185 the paper, including changing the title and figures, to emphasize the AA results. However, we 186 maintain the side-by-side comparison as we view this as a highly unique and powerful result. 187 In fact, this contrast largely addresses Major point 2 below, addressing the issue that DTT 188 189 may be only responding to a few metals and that this reviewer's analytical approach to assess this assertion should be performed on this data set, as part of this paper. 190

Changes in manuscript:

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- We added the following in the introduction section to better illustrate our purpose, page 30614 line 21, added to the end "Throughout, we compare the AA results to our previously published DTT findings (Bates et al., 2015, Fang et al., 2015b, Verma et al., 2014) to provide a clear contrast between these two commonly utilized assays to assess aerosol oxidative potential and possible associations with health endpoints."
- Changed to the title to "Oxidative Potential of Ambient Water-Soluble PM2.5 in the Southeastern United States: Contrasts in Sources and Health Associations between Ascorbic Acid (AA) and Dithiothreitol (DTT) Assays".
 - 3. Moved Figure S2 (Protocol schematics for conducting Ascorbic Acid assay) to the main text.
- 4. Modified Figure 4 by putting AA health results first. To emphasize the robustness of epi results from different models, Figure S7 (health results from the other two models) were moved to the main text and combined with Figure 4.
- 209 5. Changes were made throughout the manuscript to increase the discussions of AA results210 (see the complete manuscript with marked changes).

212 Major Points.

213 The DTT figures in the manuscript have all been shown (and discussed) previously: Figure 1b,

the DTT results in Figure 2, and Figures 3c and c are all from Verma et al. (2014), while the

215 DTT data in Figure 4 is from Bates et al. (2014). This previously published work represents

approximately half of the data in the manuscript. I appreciate that the authors want to compare

their new AA results with their old DTT results, but giving the two sets of data equal weight in

the manuscript takes away from the ascorbate findings. It also makes for a repetitious experience

for readers of Verma et al. (2014). My recommendation is to minimize the presentation of the

previous DTT figures and the discussion of the DTT results. The comparison of the DTT and AA

results is useful but could be done with a brief text discussion after each AA figure. Beyond comparisons with AA, if the authors want to present significant amounts of DTT results in the

manuscript, they should be new; see the point #2 below for some suggestions on this.

The authors should more thoroughly present and discuss the AA results, as their treatment in the

225 manuscript is often weak. For example, the authors measured hundreds of samples, but only 17

226 monthly averages are presented. Is there anything interesting to show from the time series data?

Is there anything interesting in the mass-normalized data? What do correlation plots of DTT and

228 AA rates at the various sites show? What is the average value of the (DTT rate) / (AA rate) for

each site/season; does this ratio say anything useful?

Authors' response: We have increased the discussions of AA, focusing the paper more on the AA results. As for the specific suggustions: Time series data essentially shows the same spatial and seasonal results as the average monthly data. We plotted the mass-normalized AA activity (in units of nmol/min/ μ g), see the graph below; which does not add significant new insights beyond what is already presented. DTT vs AA plots are very scattered and the ratio of DTT/AA is not very useful (this lack of correlation is imporant and discussed in the paper). Therefore, we

is not very useful (this lack of correlation is imporant and discussed in the paper). Therefore, we don't find any of the above suggested analyses worthy of more detailed discussions than what is

237 already presented in this paper.



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240 The authors have used linear regressions to assess the significance of metals and other

components in the two assays. For DTT this analysis (e.g., Figure 2) is inappropriate because (1)

two of the major contributors (Cu and Mn) have non-linear responses and (2) many of the

243 components are correlated with each other. I ranted to the authors about this in my comments to

Verma et al. (2014); see http://www.atmos-chem-phys-discuss.net/14/19625/2014/acpd-14-

19625-2014-discussion.html. Since these past comments were largely ignored, allow me rant again, both for the purposes of the current manuscript and more broadly as a statement to

247 DTT/ROS users.

We repectfully disagree with the comments that our analysis is inappropriate and that we did not already address them in a previous review (more on this below).

As an alternative to linear regressions we developed a mechanistic technique to quantify the

251 contributions of chemical species to the measured DTT (or other ROS) rate. This involves

measuring (1) concentration-response curves for each species (e.g., the rate of DTT loss as a function of copper concentration), (2) concentrations of Cu and Mn in each sample, and (3) the

DTT rate of loss in each sample. We recently compared results from linear regressions and the

255 mechanistic approach for a set of samples from Fresno, California (Charrier et al., 2015). Our

256 mechanistic approach revealed that Cu, Mn, and unknown (likely organic) species account for an

average of approximately 50%, 20%, and 30%, respectively, of the measured rates of DTT loss

258 in these samples. These percentages are approximately



Figure A. Correlations between the measured rate of DTT loss and the concentrations of Cu, Mn, and Fe in samples from Fresno, California. Black lines represent the linear regressions. Each colored line represents the calculated rate of DTT loss from the metal based on our measured concentration-response curves (Charrier and Anastasio, 2012) and the typical PM mass loading of $10 \ \mu g \ mL^{-1}$ that was used in the assay in this study. Reproduced from Figure S8 of Charrier et al. (2015).

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shown by the colored lines in Figure A. In contrast, the corresponding linear regressions for Cu,

261 Mn, an Fe show the weakest correlation for Cu (R2 = 0.40) and the strongest for Mn (R2 = 0.56).

- The Fe correlation (R2 = 0.43) is as strong as the copper correlation (and has a similar slope),
- despite the fact that copper accounts for half of DTT loss and Fe accounts for essentially none.

264 Clearly regressions cannot be trusted to identify the species responsible for DTT loss.

265 Why does this matter for the Fang et al. manuscript? Because the authors have the opportunity to

use the mechanistic approach to better assess the contributions of Cu and Mn in their samples. In

response to my first review of Verma et al. (2014), the authors calculated the contributions of

these metals for their DTT rates with the mechanistic approach. These figures show that Cu

generally makes a major contribution to the SCAPE DTT; unfortunately, the figures can only be

found in the authors' second response to my comments (http://www.atmos-chem-phys-

discuss.net/14/19625/2014/acpd-14-19625-2014-AR2.pdf) as they were not included in the final

- version of Verma et al. (2014). In a subsequent paper, Verma et al. (2015), they assessed the
- 273 contributions of transition metals towards DTT but did so using linear regressions; these
- correlations suggest that Cu and Mn are each important in only 3 of the 7 SCAPE sample sets
 examined. In contrast, the mechanistic approach results show that Cu and Mn generally dominate
- the DTT response at every site/season, although there are some problematic samples. As an

example, consider the YRK-June DTT data from Verma et al. (2015): linear regressions give R

values of 0.64, 0.53, and 0.11 for Mn, Fe, and Cu, respectively. Since its regression fell below
the R threshold, Cu was considered insignificant in these samples: the authors concluded that Mn

and organics each accounted for approximately half of the DTT response, while Cu did not

contribute. In contrast, the mechanistic approach for YRK-June (Figure B) shows that Cu

282 generally dominates the DTT response, Mn is significant, and unknown components sometimes

contribute. The measured and calculated rates for each sample in Figure B would likely agree

better if the authors measured concentration-response curves on their automated system rather
than used results from the manual runs in Charrier and Anastasio (2012). I encourage the authors

to pursue this for the revised manuscript.



Figure B. Mechanistic assessment of the transition metal contributions to the DTT response in the YRK-June (and July) samples. From the Verma et al. response (29 October 2014) to my comments to Verma et al. (2014); see http://www.atmos-chem-phys-discuss.net/14/19625/2014/acpd-14-19625-2014-AR2.pdf

It is an open question whether Cu and Mn also have non-linear responses in the AA assay. If they do, then the linear regression assessment of which species contribute to the AA result in the

290 manuscript might have problems. Since it would be a simple matter to measure the

concentration-response curves for these metals in the authors' automated system, I recommend

that they make these measurements. The authors should also use these curves to assess transition metal contributions to AA using a mechanistic approach and compare it to the regression results.

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Authors' response: We have not performed the requested analysis because we do not feel it is 295 296 informative, i.e., it does not definitively show what components are contributing to the observed 297 activity), largely because the uncertainties in the reviewers mechanistic method are very large, 298 such that it is difficult to infer definitive evidence for what species are contribution to the observed activity. The fact that this method indicates that DTT is controlled by Cu and Mn, 299 whereas we have strong evidence that this is not the case tends to support this conclusion. 300 301 Although we have addressed this before in a previous paper, we will summarize our arguments 302 again since they are stronger, in part due to the findings presented in this paper, that DTT activity 303 is not simply due to Cu and Mn. First, we summarize Figure B above by combining all the data. 304 The result is shown below as a bar/whisker plot. The summation method proposed by this reviewer, based on our measurements of water soluble Mn and Cu, significantly over-predicts 305 DTT relative to the obseved DTT, and possibly more importantly the difference is highly 306 307 variable. This is also clearly demonstrated by the r2 between measured and estimated DTT of

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308 0.22 (ie, the summation method can only explain 22% of the observed DTT variability). As we

309 have noted before, the results appear inconclusive and the method has little predictive capability.

310 (A similar conclusion is reached if one simply considers the uncertaintly in the method, when all

uncertainties are propogated). Instead of basing an analysis on laboratory experiments using purecompounds in isolation from the chemically complex mixtures found in ambient aerosols, we

prefer to be guided by ambient data based on multiple comparisons, not all of which are

correlation analyses or limited by possible nonlinear responses, ie, spatial and seasonal

distributions, and mechanistic approaches, see below. Overall, our comparisions and analysis

provides a consistent conclusion that DTT activity in our study was not driven solely by Mn and
 Cu.



It seems, however, the main point of our paper has largely been missed by this reviewer. That is, for <u>all</u> the various comparisons outlined here, there are distinct contrasts between the DTT and AA assays, and the two assays are not themselves well correlated. If DTT was only responsive to Cu and Mn, (like AA), the stark contrasts between DTT and AA in the ambient data outlined here would not be expected. Again, we feel this demonstrates the strength of our papers format of a side-by-side comparison between these assays.

In further support of our approach, next we summarize the evidence that DTT is driven by more than Cu and Mn.

1) The major emission source found to be most influential from our two different source

apportionment models (PMF and CMB-E) is

biomass burning (Figure 3). It is known that

336 biomass burning is not a large source for Cu and Mn, and indeed our study shows that only 16% of the Mn and 0% of Cu is associated with this source [Fang et al., 2015a]. Biomass burning, 337 however, is a known large source of OA, specifically HULIS, which we (and others) have shown 338 is related to DTT activity and which contains compounds known to be DTT active (eg., 339 340 quinones, , hydroxyquinones, etc.). Furthermore, in this paper we show AA is mostly associated with these metals and biomass burning is not a source for AA. The seasonal differences in DTT 341 342 and AA, also reported in this paper, provide further support; where DTT activites are higher in 343 winter at some sites, AA activities were not (AA actually has the opposite trend, being higher in 344 summer), consistent with known trends in biomass burning emissions (ie, higher in winter) and 345 confirmed by our tracers. Likewise, AA source apportionment indicated it was more associated 346 with traffic sources, which is supported by the spatial distributions; AA was observed to be higher at urban vs rural sites, and amongst the urban sites higher at sites closer to traffic 347 emissions. DTT does not show similar spatial trends. Thus, although the source approtionment 348 may be affected by co-variabilities and non-linear responses, the season and spatial trends are not, 349 350 and the two indepenant analyses are consistent.

2) Another important point discussed in this paper is that DTT was highly correlated with $PM_{2.5}$ 351 352 mass (r=0.49-0.88 [Fang et al., 2015c]). Mn and Cu do not correlate with PM_{2.5} mass at most 353 of times, (r=-0.18-0.66, two exceptions are at GT and RS winter time, r>0.75 for Mn), which 354 is not surprising as they are small mass fractions of PM_{2.5}. If Cu and Mn were the main drivers of DTT as well, one wouldn't expect to see the high correlations between DTT and 355 PM mass. This is noteworthy since it is consistent with our epi comparisons between DTT 356 357 and AA, and the many studies showing linkages between $PM_{2.5}$ mass and health. DTT is correlated with PM2.5 and DTT shows associations with some health endpoints, AA is not 358 correlated with PM_{2.5} mass and shows less assocation with health endpoints. 359 3) This reivewer perfers a mechanistic analysis versus correlations. We have also done this and 360

the results are consistent with the correlation analyses. A mechanistic approach that 361 362 quantified the contributions of OA vs metals to DTT has been presented [Verma et al., 2015a]. In that study we passed the water-soluble extracts, in which both water-soluble DTT 363 activity and metals concentrations had been measured, through a C-18 column to separate 364 hydrophilic and hydrophonic components. Both the DTT activity and metals concentrations 365 were measured in the fraction that passed through the column (hydrophilic fraction). The 366 hydrophobic fraction can be determined by the difference between total and hydrophibic. We 367 found that hydrophobic DTT activity made up a large fraction of total water-soluble DTT 368 activity and Cu and Mn were mostly excluded from the hydropholic fraction. These results 369 indicated that the DTT activity of the water-soluble hydrophobic fraction was not driven by 370 Cu/Mn, but another class of redox-active species, most probably organic compounds 371 associated with HULIS. From this analyses we reported that roughly 60% of the DTT 372 activity in that study was associated with the OA and 40% metals. As OA from biomass 373 burning is a major source for HULIS, this is consistent with the various source 374 apportionment and correlation analysis discussed above. Finally, we also did a mass balance 375 376 analaysis, much like the reviewer has done, and found that including the OA (ie, hydrophobic DTT activity) provided better closure. 377

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4) As further proof that DTT is not soley driven by metals we have performed a traffic
tunnel/chamber study, which has shown high DTT activity for ambient aerosol containing no
metals. In yet-to-be-published work we performed experiments on sample air that was drawn
from a highly trafficed vehicle tunnel. Tunnel air could be analyzed directly or modified. In
one experiment, the sample air was filtered to remove all aerosols (which includes all
metals). The sample was then directed to a smog chamber and explosed to lights, resulting in
SOA from the VOCs in the tunnel. This SOA was highly DTT active, with levels comparable

- to the hyrdophobic fraction of the ambient aerosols observed in SCAPE.
- 387

388 In summary, although this reviewer's point is well taken, that correlation is not causation, the

totallity of our uniquely large data set strongly point to the role of both OA and metals to the

390 DTT activity of water soluble components. It includes analyses that are not subject to the

limitations pointed out (correlation analysis and possible nonlinear responses). The uniquestrength of our data is that it is large and includes long time series (ie, seasonal data), and data

from multiple sites (ie, spatial data). This is in contrast to the reviewers work, from which they

conclude that DTT is mainly driven by Cu and Mn. We have also shown above that the analysis 394 proposed, does not provide a definitive indication that DT activity is driven solely by Mn and 395 Cu, for our data set. And finally, based on this paper comparing DTT and AA, the Verma et al., 396 397 (2015) paper and our tunnel data, we conclude that the approach used by this reviewer does not provide an informative way to assess the contributions of aerosol species to the water-soluble 398 DTT activity of ambient particles measured in our study. Finally, some time back we attempted 399 the mechanistic approach, developing a calibration for AA response to pure Cu solutions made 400 from copper sulfate. Although only exploratory in nature, we found the calibration curve to be 401 402 somewhat unstable and the predicted AA, based on the curve and our measured water soluble Cu, greatly under predicted the AA activity relative to observed (10% of observed). For all of 403 these reasons, we chose not to include this reviewer's mechanistic approach in the analyses of 404 405 our AA data set.

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407 4. Backcast estimates of AA and DTT activities and the epidemiological analyses. A discussion

of the uncertainties in the backcast estimates of the AA response is needed. The uncertainties 408

must be very large, as illustrated for DTT in Figure 1 of Bates et al. (2015). The equivalent 409

figure for AA should be shown in the manuscript. As part of the discussion, how can the 410 backcast uncertainties of AA (and DTT) activity be enormous, but the 95% confidence intervals 411

412 around the RR data points in Figure 4 be quite small. Do the Figure 4 CIs include the full

413 backcast uncertainties? How are these propagated?

Authors' response: The CIs in the figure are for the epi results and do not include uncertainty in 414 the backcast DTT or AA. See comments to minor point 3 below. Our argument is that the AA 415 and DTT models do roughly equally well when compared to the ambient data (similar r). How 416 well they predict actual historical DTT and AA is not known, but we have no reason to believe 417 that AA is predicted much worse than DTT. In fact one might expect the opposite as DTT is 418 driven by more sources than AA. In any case, the point is that DTT shows health associations 419 and the AA does not. This is an important conclusion despite the high uncertainties with the 420 421 methods, and is consistent with the contrasts described throughout the paper. It again stresses the value of the side-by-side comparison. 422 423

424 Changes in manuscript: No changes.

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426 Since AA is dominated by Cu, and there are enormous uncertainties in the backcast estimates of AA (and DTT) activity, it would be interesting to do the epidemiological modeling using 427

measured particulate Cu rather than the predicted AA response. This might show a significant 428

correlation with the health endpoints. Are there historical data in Atlanta that could be used for 429

this? If this epidemiological analysis could be done relatively easily, I encourage the authors to 430

431 include it in the current manuscript. If not, I hope to see it in a future manuscript.

432 Authors' response: We have considered this. Unfortunately, a sufficient water-soluble Cu data

433 does (ie. Sufficient number of data above LOD) not exist for the time period of the epi study.

Thus, running the epi with just water-soluble Cu would still require a backcast prediction. 434

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436 Changes in manuscript: No change was made.

Minor Points.

441 442	analogous to the DTT nomenclature of Verma et al. (2015). There is the same issue for DTT in the manuscript.
443 444 445	Authors' response: Replaced DTT and AA activity with DTTv and AAv throughout the paper. Also used DTT_v^e and AA_v^e for backcast estimates of DTTv and AAv.
446 447	2. p.30630. The authors should include the results for the other health outcomes (COPD, pneumonia, IHD) in Table S4, as the comparison with asthma and CHF would be interesting.
448 449 450	Authors' response: We agree and added the health outcomes of COPD, pneumonia, and IHD to Table S4.
451 452 453 454	Changes in manuscript: added the data of health risk ratios of COPD, IHD, and pneumonia to Table S4. In the main text, page 30630 line 12: changed "(results not presented)" to "(results given in Table S4)".
455 456 457 458 459	3. p. 30631. The authors conclude that "For the region investigated in this study, the DTT assay was a more comprehensive multi-pollutant ROS (or oxidative potential) indicator than the AA assay making DTT a potentially valuable parameter to include in future PM health-related studies." Given the very small differences in the RRs for DTT and AA, and the very large uncertainties in their backcast estimates, this conclusion is far too strong.
460 461 462 463 464 465	Authors' response: We disagree when one considers all aspects of the comparisons laid out in this paper. Admittedly, the back-cast prediction of DTT and AA activities has large uncertainties, but again it is the comparisons that are insightful; DTT shows statistically significant associations (at 95% confidence level), AA does not. And this is true for all regression models we used to backcast DTT and AA. We have now added these results to Figure 4 to emphasis the point.
466 467 468 469 470 471	Additionally, previous studies has shown good correlation between DTT and biomarkers such as hemeoxygenase expression in cells and exhaled nitric oxide fraction [<i>Delfino et al.</i> , 2013; <i>Li et al.</i> , 2003]. We believe that DTT assay is a valid accellular assay for studying the ROS-generation <i>in vivo</i> and would be valuable to include in future PM health-related studies.
472 473 474 475 476	Changes in manuscript: Page 30630 line 18, edited as "The same results were found for estimates based on the two other regressions [(see Figure 5(b) and (c)]), suggesting that the null relationship of AA_v^e and positive association of DTT_v^e with these health outcomes are to some extent robust, despite the high uncertainties from the back-cast models."

1. "AA" is used to represent the volume-normalized rate of AA loss, but of course it's also the name of AA itself, which is confusing. Better to use something like "AAv" for the rate,

4. Finally, I want to apologize to the authors for taking so long to complete this review.

- 478
- 479
- 480 Anonymous Referee #3
- 481
- 482 Summary

This manuscript uses a number of important methods to bridge the continuing gap between detailed composition information and the association between PM and negative health effects. The DTT and AA assays were run on PM extracted from a number of locations across two year in the southeast United States. Correlations were made between the DTT and AA assays and PMF and CMB modeling of sources and with ED admissions from long term epidemiological data. On the whole I feel the paper has a lot of strong characteristics, particularly the new assay

methodologies and their connection with source modeling. I think the paper overall is fairly

- 490 strong and should be accepted with minor revisions.
- 491

492 Major Comments

One concern is that there is a lot of detail and information in the methods section, but that the results and discussion section feels light on new results. As noted by another reviewer, a large amount of the DTT data has been shown before in prior publications. Though comparisons with prior data are good, it feels a little lik a second bit at the apple for the DTT data. A revised manuscript that shows more data from the AA results would be compelling, though perhaps due to the strong Cu response there was concern that the AA results were not strong enough on their own? I rarely suggest making a paper longer, but at 18 pages with only 4 of results and discussion that could really be expanded and strengthened.

discussion that could really be expanded and strengthened.

Authors' response: Since this work has combined field campaign, air quality modeling and
 epidemiologic analysis, it involves extensive methodological descriptions. We have eliminated
 the replicate method descriptions, now providing just references, and have moved them to the
 supplemental materials. These include the method description of DTT assay from Fang et al.
 (2015b) and water-soluble elements from Fang et al. (2015a).

We have also revised the paper, putting more emphasis on the AA results. Also see our 2^{nd} response to the 2^{nd} reviewer's comments.

511 Changes in manuscript: Moved the method description of DTT assay (page 30615 section 512 2.2.1) and a large fraction of "water-soluble elements" (page 30620 section 2.3.2) to the 513 supplement and added text emphasizing the AA results.

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510

The extensive literature citations are commendable and the introduction very nicely sets up the need for this work. Along the lines of the previous comment, at times I felt like the new findings

- from this study got lost in the sea of references to this group of authors previous papers,
- 518 particularly those from the last 2 years and Balachandran et al. 2012. A revised manuscript that

519 more clearly delineates the findings of this study from prior work will help readers understand 520 the findings of this work. Though the data is admittedly abundant and at times dense, the results 521 section feels like a list of numbers in a few paragraphs. Rewording to bring the science out from 522 behind these numbers would be helpful.

Authors' response: We have eliminated repetitive references to previous publications and havereworded most sections to emphasize the results from AA.

525

In Figure 3 the amount of AA response from brake/tire wear being 44% is notable. I worry that 526 527 since as the authors note that these are mechanically generated particles that they are likely solid and smaller than the 0.45 micron filter used in the methods. The still solid metals might only be 528 dissolved when nitric acid is introduced prior to the XRF analysis. If so this would seem to 529 suggest that this is not truly a water soluble, but rather an insoluble source that becomes soluble 530 531 during analysis. Overall the possibility of solid particles less than the 0.45 um filter should be 532 addressed in more detail. If this concern is plausible then some of the phrasing in the manuscript (and potentially even the title) should be qualified since much of the AA response might be from 533 insoluble material. 534

535 Authors' response: If the particles are formed mechanically, they are not expected to be smaller 536 than 0.45 µm in significant quantities. It is established that mechanically generated aerosol are 537 mainly found in the coarse mode, but which can extend down to ~1um (eg, see Seinfeld and Pandis). Thus, some fraction of these coarse particles will be included in PM2.5, but very few 538 below 0.45 um. (Note, we have measured the size distribution of the AA activity and there is 539 practically no contribution to the AA activity from particles below 0.45 um). Thus, we expect 540 few solid particle to interfere with our measurement. Regarding the nitric acid: Nitric acid is 541 542 added to the extract after the extract is filtered, so it will not add to the water soluble metals 543 signal if there are insoluble particle in the extract (ie, particles below 0.45 um).

544

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545 Changes in manuscript: No changes.

547 AA is used as a surrogate for a more complex lung fluid and found to only really correlate with 548 Cu. Are the authors concerned that perhaps this more simplistic model fluid is not capturing 549 sensitivity to other species that perhaps a more complex simulated fluid might? A little bit of 550 further discussion on whether AA is really a great choice would be helpful, a otherwise it might 551 not be sensitive enough to mimic the responses in the lung that are observed from other types of 552 studies.

553

Authors' response: Yes, the AA assay used in our work (as stated in page 30617 line 14) is a more simplistic model to a synthetic respiratory tract lining fluid model (RTLF), which contains, besides AA, urate and reduced glutathione (GSH). This ascorbate-only method has been used by other studies [*Ayres et al.*, 2008; *Mudway et al.*, 2005] to provide an alternative high throughput method to synthetic RTLF. In their work, AA-only method was performed to characterize the metal dependence of oxidative reactions, mimicking the nature of particles-antioxidant (i.e. AA) interactions at the air lung interface.

Changes in manuscript: page 30617 line 14, changed "The method in this study was based on 562 an ascorbate-only model (Mudway et al., 2005; Ayres et al., 2008) that is a simplified approach 563 to a synthetic respiratory ..." To "The method in this study was based on an ascorbate-only 564 model (Mudway et al., 2005; Ayres et al., 2008) that is a simplified and alternative high 565 throughput approach to a synthetic respiratory ... " 566 567 Minor Comments/Concerns 568 LWCC is defined on both 30617 and 30619 569 Authors' response: deleted the 2nd expansion of LWCC. 570 571 Changes in manuscript: Page 30619 line 19: changed "Liquid Wave-guide Capillary Cell" to 572 "LWCC". 573 574 575 700 nm was chosen as the background, but some things in the atmosphere can absorb at that wavelength. Washenfelder et al. 2015 GRL showed episodic variation of absorption for the 4 576 wavelength of aerosol absorption they monitored (highest was 530 nm). Have the authors done 577 checks with full absorption spectra to determine if 700 nm is truly a safe choice for background? 578 This would be particularly important when biomass is playing such an important role in both 579 580 absorption and DTT assay activity. Authors' response: We developed the method used to measure Brown Carbon [Hecobian et al., 581 2010] and made the measurements (ie, it is our data) and can say definitively that there is no 582 absorption at 700 nm from other aerosol components. 583 584 585 Changes in manuscript: No changes. 586 587 As a curiosity is a there a reason that XRF was used and not ICP-MS? Authors' response: It was the method available and also easily adaptable to handle the large 588 589 number of filters we had to analyze. To send out the hundreds of samples for analysis by ICP-MS would have lead to exorbitant costs (ie, over \$100/sample) 590 591 Changes in manuscript: No changes. 592 593 594 This might just be a lack of knowledge, but ammonium bisulfate is a class listed for one of the 595 models. How is that determined and separated from ammonium sulfate? Is that from ISOROPPIA or some other thermodynamic model? 596 Authors' response: The separation of ammonium sulfate and ammonium bisulfate is based on 597 the differences in their source profiles, which were determined from an ensemble average of 598 599 different CMB profiles and CMAQ results. The main distinguishing drivers in the source profiles are the average mass ratio of ammonium/sulfate. A thermodynamic model (ie, ISORROPIA) is 600 not used. 601

603 Changes in manuscript: No changes.

- 604 Section 2.3.3 Though referencing the other work is important, I am a bit suspicious of using PM 605 mass concentrations estimated from the sum of chemical component from Hi-Vol samples. More 606 detail on how this has been justified (even one sentence), beyond just referencing citations 607 would appease that concern. 608 Authors' response: Since total PM_{2.5} measurement was not available for the RS and GT site, we 609 610 had to use the summation method. Added the statement below to clarify. 611 612 Changes in manuscript: Page 30620 line 20: changed "For the RS and GT sites, the PM mass concentrations were estimated from the sum of chemical components analyzed on the same Hi-613 Vol" to "For the RS and GT sites, since PM2.5 mass were not available, the PM mass 614 615 concentrations were estimated from the sum of chemical components analyzed on the same Hi-616 Vol filters (Verma et al., 2014) (Details in the Supplement)." 617 618 Are the %'s (filter sampling, etc.) chosen for propagating uncertainty on page 30632 from a 619 reference or chosen arbitrarily? I was not clear from the text where they came from. Authors' response: Our previous work [Fang et al., 2015b] has discussed in details the 620 uncertainties from the method, including analytical uncertainty (2-15%), calibration uncertainty 621 (1-6%), and blank variability. We have found that the overall precisions from co-located 622 comparisons (1σ /slope) are 6-20%. From this, we estimated the specific uncertainties for filter 623 sampling and extraction to be 5%. 624 625 Changes in manuscript: Page 30621 line 16, changed "which were obtained by propagating the 626 uncertainties from filter sampling (5%), extraction (5%)" to "which were obtained by 627 propagating the uncertainties from filter sampling (assumed to be 5%), extraction (assumed to be 628 629 5%)". 630 631 Page 30630: "the model did not observed" should be "the model did not observe" 632 Authors' response: edited as suggested. 633 634 **Changes in manuscript:** page 30630 line 31: changed "the model did not observed" to "the 635 model did not observe". 636 637
- 638 Ayres, J. G., P. Borm, F. R. Cassee, V. Castranova, K. Donaldson, A. Ghio, R. M. Harrison, R. Hider, F. Kelly,
- and I. M. Kooter (2008), Evaluating the toxicity of airborne particulate matter and nanoparticles by
- measuring oxidative stress potential-a workshop report and consensus statement, *Inhalation toxicology*,
 20(1), 75-99.

- Delfino, R. J., N. Staimer, T. Tjoa, D. L. Gillen, J. J. Schauer, and M. M. Shafer (2013), Airway inflammation
 and oxidative potential of air pollutant particles in a pediatric asthma panel, *J Expos Sci Environ*
- 644 *Epidemiol, 23,* 466-473.
- Fang, T., H. Guo, V. Verma, R. E. Peltier, and R. J. Weber (2015a), PM2.5 water-soluble elements in the
- 646 southeastern United States: automated analytical method development, spatiotemporal distributions,
- source apportionment, and implications for heath studies, *Atmos. Chem. Phys. Discuss.*, *15*(12), 1718917227.
- Fang, T., H. Guo, V. Verma, R. E. Peltier, and R. J. Weber (2015b), PM2.5 water-soluble elements in the
- southeastern United States: automated analytical method development, spatiotemporal distributions,
 source apportionment, and implications for heath studies, *Atmos. Chem. Phys.*, *15*(20), 11667-11682.
- Fang, T., V. Verma, H. Guo, L. E. King, E. S. Edgerton, and R. J. Weber (2015c), A semi-automated system
- for quantifying the oxidative potential of ambient particles in aqueous extracts using the dithiothreitol
- (DTT) assay: results from the Southeastern Center for Air Pollution and Epidemiology (SCAPE), Atmos.
 Meas. Tech., 8(1), 471-482.
- Hecobian, A., X. Zhang, M. Zheng, N. Frank, E. S. Edgerton, and R. J. Weber (2010), Water-Soluble
- Organic Aerosol material and the light-absorption characteristics of aqueous extracts measured over the
 Southeastern United States, *Atmos. Chem. Phys.*, *10*(13), 5965-5977.
- Li, N., C. Sioutas, A. K. Cho, D. Schmitz, C. Misra, J. Sempf, M. Wang, T. Oberley, J. Froines, and A. Nel
- 660 (2003), Ultrafine particulate pollutants induce oxidative stress and mitochondrial damage,
- 661 Environmental Health Perspectives, 111(4), 455-460.
- 662 Miljevic, B., F. Hedayat, S. Stevanovic, K. E. Fairfull-Smith, S. E. Bottle, and Z. D. Ristovski (2014), To
- Sonicate or Not to Sonicate PM Filters: Reactive Oxygen Species Generation Upon Ultrasonic Irradiation,
 Aerosol Science and Technology, 48(12), 1276-1284.
- 665 Mudway, I., S. Duggan, C. Venkataraman, G. Habib, F. Kelly, and J. Grigg (2005), Combustion of dried
- animal dung as biofuel results in the generation of highly redox active fine particulates, *Particle and Fibre Toxicology*, 2(1), 6.
- 668 Verma, V., T. Fang, L. Xu, R. E. Peltier, A. G. Russell, N. L. Ng, and R. J. Weber (2015a), Organic aerosols
- associated with the generation of Reactive Oxygen Species (ROS) by water-soluble PM2.5,
- 670 Environmental Science & Technology, 49(7), 4646-4656.
- 671 Verma, V., Y. Wang, R. El-Afifi, T. Fang, J. Rowland, A. G. Russell, and R. J. Weber (2015b), Fractionating
- 672 ambient humic-like substances (HULIS) for their reactive oxygen species activity Assessing the
- 673 importance of quinones and atmospheric aging, *Atmospheric Environment*, *120*, 351-359.
- 674

676	Oxidative Potential of Ambient Water-Soluble PM_{2.5}
677	Measured by Dithiothreitol (DTT) and Ascorbic Acid (AA)
678	Assays-in the Southeastern United States: Contrasts in
679	Sources and Health Associations between Ascorbic Acid
680	(AA) and Dithiothreitol (DTT) Assays
681	
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707 Abstract

706

The ability of certain components of particulate matter to induce oxidative stress through catalytic 708 generation of reactive oxygen species (ROS) in vivo may be one mechanism accounting for 709 710 observed linkages between ambient aerosols and adverse health outcomes. A variety of assays 711 have been used to measure this so-called aerosol oxidative potential. We developed a semi-712 automated system to quantify oxidative potential of filter aqueous extracts utilizing the 713 dithiothreitol (DTT) assay and have recently developed a similar semi-automated system using the ascorbic acid (AA) assay. Approximately 500 PM2.5 filter samples collected in contrasting 714 715 locations in the Southeastern US were analyzed for a host of aerosol species, along with AA and 716 DTT activities. Here we present a detailed contrast in findings from these two assays. using both 717 assays. We found that wWater-soluble DTT activity on a per air volume basis was more spatially 718 uniform than water soluble AA activity. DTT-AA activity was higher in summer/fall winter than 719 in winter-summer/fall, with highest levels near highly trafficked highways, whereas AA-DTT 720 activity was higher in winter summer/fall compared to summer/fall and more spatially 721 homogeneous-winter, with highest levels near highly trafficked highways. AA activity was nearly exclusively correlated with water-soluble Cu (r = 0.70-0.94 at most sites), whereas DTT activity 722 was correlated with organic and metal species, whereas AA activity was correlated with water-723 724 soluble metals (especially water soluble Cu, $r = 0.70 \cdot 0.91$ at most sites). Source apportionment 725 models, Positive Matrix Factorization (PMF) and a Chemical Mass Balance Method with 726 ensemble-averaged source impact profiles (CMB-E), suggest a strong contribution from secondary processes (e.g., organic aerosol oxidation or metal mobilization by formation of an aqueous 727 728 particle with secondary acids) and traffic emissions to both AA DTT and DTT AA activities in 729 urban Atlanta. Biomass burning was a large source for DTT activity, but insignificant for AA. AA 730 activity DTT activity was well-not correlated with $PM_{2.5}$ mass ($r = 0.49 \cdot 0.86$ across sites/seasons), 731 while DTT activity AA activity did not co-varied strongly with mass (r = 0.49-0.86 across 732 sites/seasons). A-Various linear models was-were developed to estimate AA DTT and DTT AA 733 activities for the central Atlanta Jefferson Street site, based on the CMB-E sources-that-are 734 statistically significant with positive coefficients. The models was were then used to estimate daily 735 oxidative potential at this site over the period-1998-2009 period. Time-series epidemiological analyses were conducted to assess daily emergency department (ED) visits data for the five-county 736 737 Atlanta metropolitan area based on the estimated 10-year backcast oxidative potential. -Results 738 suggest that estimated AA activity was not statistically associated with any tested health outcomes, 739 while DTT activity was associated with ED visits for both asthma/wheeze and congestive heart 740 failure, while AA activity was not linked to any health outcomes. The findings point to the importance of both organic components and transition metals from biomass burning and mobile 741 sources to adverse health outcomes in this region. 742

- 744 Key Words: water-soluble DTT activity; water-soluble AA activity; oxidative potential;
- 745 source apportionment; health associations; cardiovascular; respiratory

746 I

747 1. Introduction

748	Studies have linked exposure to fine particulate matter $(PM_{2.5})$ with increased respiratory	
749	(Harkema et al., 2004; Aust, 2002; Schaumann et al., 2004) and cardiovascular (Pope et al.,	
750	2004; Samet et al., 2000) diseases. PM _{2.5} consists of a wide range of chemical components of	
751	potentially varying toxicity, implying that PM _{2.5} is not an ideal air quality metric for assessing	
752	health impacts. For example, components such as ammonium, sulfate, nitrate, chloride, and some	
753	chemical fraction of mineral dust, may be more benign than transition metals (Gasser et al.,	
754	2009; Kodavanti et al., 2005; Akhtar et al., 2010), black carbon (or elemental carbon and	
755	associated species) (Kleinman et al., 2007; Brunekreef et al., 1997), polycyclic aromatic	
756	hydrocarbons (PAHs) (Lundstedt et al., 2007; Burchiel et al., 2005), and other specific organics	
757	species (Nel et al., 2001). Although a small mass fraction of PM2.5, these components may could	
758	play a disproportionally large role in the overall adverse health effects of PM _{2.5} . In contrast,	
759	Harrison et al. (2000) notes that the consistency of epidemiological associations between PM2.5	
760	mass and health effects across regions of widely varying aerosol sources implies that individual	
761	chemical components comprising small mass fractions of PM2.5-cannot be responsible for	
762	observed PM2.5-health associations. A comprehensive set of mechanisms explaining the observed	
763	linkage between PM2.5 mass and adverse health effects has not been established, but it has been	
764	hypothesized that one possible contributing mechanism physiological route is a particle's ability	
765	to induce oxidative stress via catalytic generation of reactive oxygen species (ROS), in vivo. A	Formatted: Font: Italic
766	number of studies have associated particle oxidative capacity with PM toxicity (Donaldson et al.,	
767	2005; Nel, 2005; Shi et al., 2003; Zielinski et al., 1999), but without available large databases of	

769	have not been possible.	
770	A number of different assays have been developed to quantify the oxidative potential of PM	
771	samples (Zomer et al., 2011; Mudway et al., 2011; Ayres et al., 2008; Jung et al., 2006; Cho et	
772	al., 2005; Mudway et al., 2005; Venkatachari et al., 2005). Two commonly used approaches are	
773	the dithiothreitol (DTT) assay (Cho et al., 2005) and here describe its adaptation to the ascorbic	
774	acid (AA) assay (Ayres et al., 2008; Mudway et al., 2005). It may be expected that these two	
775	different assays respond to different aerosol components and are linked to different health	
776	endpoints. The AA assay has been shown to be most sensitive to transition metals (Janssen et al.,	
777	2014; Strak et al., 2012; DiStefano et al., 2009; Künzli et al., 2006) but quinone compounds may	
778	react with AA as well (Roginsky et al., 1999). For the DTT assay, identified DTT-active PM	
779	components are organic species, including water-soluble organic carbon (WSOC) (Verma et al.,	
780	2009; Cho et al., 2005), or of increasing specificity, HUmic-LIke Substances (HULIS) (Verma et	
781	al., 2012; Lin and Yu, 2011), and quinones (Chung et al., 2006; Kumagai et al., 2002) (a	
782	component of HULIS). Other studies, however, have emphasized the role of transition metals,	
783	such as Cu and Mn (Vejerano et al., 2015; Charrier and Anastasio, 2012). Some differences	
784	between studies may arise due to differing source characteristics of the specific regions studied.	
785	Both assays involve incubating the anti-oxidant (DTT or AA) with filter aqueous extracts of	
786	PM _{2.5} at a controlled temperature (37 °C) and pH (7.4), and measuring the depletion of the	
787	antioxidant over time, typically detected as a decrease in absorbance light absorption at a certain	
788	wavelengths of(-412 and 265 nm for DTT and AA, respectively). The antioxidant loss rate is	
789	interpreted as a measure of the ability of aerosol redox-active species to catalytically transfer	
790	electrons from DTT or AA to oxygen (O2). DTT can be considered a chemical surrogate to	

ambient aerosol ROS, large population-based epidemiologic studies of $PM_{2.5}$ oxidative potential

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Field Code Changed

791	cellular reductants, such as NADH or NADPH, which reduces O_2 to superoxide anion (O_2^{-}) and
792	induces oxidative stress (Kumagai et al., 2002). Unlike DTT, AA is a physiological antioxidant
793	in lung lining fluid, which prevents the oxidation of lipids and proteins (Valko et al., 2005).
794	Asthmatic patients have markedly decreased concentration of AA in lung lining fluid compared
795	to healthy control subjects (Kelly et al., 1999). Therefore, the in vitro oxidation of these two
796	antioxidants by PM might represent the interaction of PM with biological antioxidants within the
797	bodyin vivo leading to the induction of oxidative stress and ultimately adverse health effects.
798	Among the various available methods for measuring oxidative potential, these two assays are
799	relatively straightforward and reproducible, allowing high throughput routine measurements and
800	the generation of large data sets for exploring links between aerosol components and health
801	through epidemiology, or also as an initial screening step for identifying different redox
802	components for more detailed cell or animal studies (Ayres et al., 2008).
803	We recently developed a semi-automated system (Fang et al., 2015b) to measure DTT activity
804	and here describe its adaption to the AA assay. Utilizing our automated analytical system, we
805	measured the water-soluble oxidative potential of over 500 filter samples collected as part of the
806	Southeastern Center for Air Pollution & Epidemiology (SCAPE) study. Although in-soluble
807	components are important, since there is no current standard protocol for measuring the water-
808	insoluble oxidative potential, we focus solely on the water-soluble AA and DTT activities. We
809	evaluate and compare these two assays in order to identify specific aerosol components the AA
810	assay is responsive to. We perform a source apportionment analysis and assess these results
811	through observed AA activity seasonal and spatial variability. AA source profiles are used to
812	generate a model that estimates AA activities, which is then used to backcast AA levels over the
813	past 10 years for use in a time-series epidemiological analysis in the Atlanta metropolitan area.

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814	Throughout, we compare the AA results to our previously published DTT findings (Bates et al.,
815	2015, Fang et al., 2015b, Verma et al., 2014) to provide a contrast between these two commonly
816	utilized assays to assess aerosol water-soluble oxidative potential and possible associations with
817	health endpoints.
818	
819	quantify oxidative potential of PM samples utilizing the dithiothreitol (DTT) assay (Cho et al.,
820	2005) and here describe its adaptation to the ascorbie acid (AA) assay (Ayres et al., 2008;
821	Mudway et al., 2005). Among the various available methods for measuring oxidative potential,
822	these two assays are relatively straightforward and reproducible, allowing high throughput
823	routine measurements and the generation of large data sets for exploring links between aerosol
824	components and health through epidemiology, or also as an initial screening step for identifying
825	different redox components for more detailed cell or animal studies (Ayres et al., 2008). Janssen
826	et al. (2015) investigated the two assays in a panel study (n=31) and found significant
827	associations between both assays and nasal and airway inflammation. Both assays involve
828	incubating the anti-oxidant (DTT or AA) with filter aqueous extracts of $PM_{2.5}$ at a controlled
829	temperature (37 $^{\circ}$ C) and pH (7.4), and measuring the depletion of the antioxidant over time,
830	typically detected as a decrease in absorbance at wavelengths of 412 and 265 nm for DTT and
831	A.A., respectively. The antioxidant loss rate is interpreted as a measure of the ability of acrosol
832	redox-active species to catalytically transfer electrons from DTT or AA to oxygen (O2). DTT can
833	be considered a chemical surrogate to cellular reductants, such as NADH or NADPH, which
834	reduces Θ_2 to superoxide anion (Θ_2) and induces oxidative stress (Kumagai et al., 2002). Unlike
835	DTT, AA is a physiological antioxidant in lung lining fluid, which prevents the oxidation of
836	lipids and proteins (Valko et al., 2005). Asthmatic patients have markedly decreased

837	concentration of AA in lung lining fluid compared to healthy control subjects (Kelly et al.,
838	1999). Therefore, the in vitro oxidation of these two antioxidants by PM might represent the
839	interaction of PM with biological antioxidants within the body leading to the induction of
840	oxidative stress and ultimately adverse health effects.
841	It may be expected, however, that these two different assays respond to different aerosol
842	components and are linked to different health endpoints. The AA assay has been shown to be
843	most sensitive to transition metals (Janssen et al., 2014; Strak et al., 2012; DiStefano et al., 2009;
844	Künzli et al., 2006) but quinone compounds may react with AA as well (Roginsky et al., 1999).
845	For the DTT assay, identified DTT active PM components are organic species, including water-
846	soluble organic carbon (WSOC) (Verma et al., 2009a; Cho et al., 2005), or of increasing
847	specificity, HUmic-Like Substances (HULIS) (Verma et al., 2012; Lin and Yu, 2011), and
848	quinones (Chung et al., 2006; Kumagai et al., 2002) (a component of HULIS). Other studies,
849	however, have emphasized the role of transition metals, such as Cu and Mn (Vejerano et al.,
850	2015; Charrier and Anastasio, 2012). Some differences between studies may arise due to
851	differing source characteristics of the specific regions studied.
852	Utilizing our automated analytical system (Fang et al., 2015b), we measured the water soluble
853	oxidative potential of over 500 filter samples collected as part of the Southeastern Center for Air
854	Pollution & Epidemiology (SCAPE) study using the DTT and AA assays. Here we evaluate and
855	compare the two assays in order to identify the sources and specific aerosol components the
856	assays respond to. We use the sources to generate a model that estimates DTT and AA activities
857	and compare these estimated levels to the measured data. The model is then used to backcast
858	DTT and AA levels over the past 10 years for use in a time series epidemiological analysis in the
859	Atlanta metropolitan area.

860 2. Methods

861 **2.1. Sampling**

862 Sampling methods have been described in detail elsewherein our previous publications (Fang et al., 2015b; Verma et al., 2014). In brief, PM_{2.5} (quartz filters, Pallflex® Tissuquartz[™], 8 × 10 863 inches) was sampled at seven locations in the Southeastern US, with different source 864 characteristics, using two sets of high-volume samplers (Hi-Vol) (Thermo Anderson, flow rate 865 normally 1.13 m³ min⁻¹). Sampling in the metropolitan Atlanta area was carried out from June 866 2012 through March 2013 (noon - 11 a.m., 23 hours) and involved paired-sites with one Hi-Vol 867 868 sampler fixed at an urban background site (Jefferson Street, referred as JST) whilst the other sampler was deployed at three other sites on a monthly basis, and at least twice during different 869 seasons. These three sites were: a rural site (Yorkville, YRK), a road-side site (RS, adjacent to 870 871 the interstate highway I75/85), and a near-road site (GT, 840 m from the RS site). Following sampling in Atlanta, the two samplers were moved to Birmingham, AL (BHM, within a few 872 873 kilometers of significant transportation and industrial sources) and Centerville, AL (CTR, surrounded by forests and a lightly traveled county road) for a month of sampling in June-July 874 875 2013, followed by an-one-month August sampling at East St. Louis, IL, an urban residential/light 876 commercial area about 3 km east of the central business district of St. Louis, MO (Sauvain et al., 2008). Finally, a GT-RS pair was conducted in September 2013. A table providing the sampling 877 schedule and a map can be found in the supporting materials (Table S1 & Fig. S1). JST, YRK, 878 879 BHM, and CTR are all part of the Southeastern Aerosol Research and Characterization Study 880 (SEARCH) network sites (Hansen et al., 2003). Collected samples were immediately wrapped in prebaked aluminum foil and stored at -18° C until analyzed. DTT, water-soluble organic carbon, 881 882 and brown carbon analyses on the filters were conducted within a year of sample collection,

883 water-soluble elements were within a year and half, and AA measurements were conducted

884 within two years of sample collection.

885 2.2. Oxidative potential measurement

886 **2.2.1.** Semi-automated system for measuring water-soluble DTT activity

Filter extraction: Three punches of the collected Hi Vol filter (5.07 cm² each) were extracted in 887 888 15 mL of deionized (DI) water (> 18 M Ω cm⁻¹) in a sterile polypropylene centrifuge tube (VWR 889 International LLC, Suwanee, GA, USA) by sonication using an Ultrasonic Cleanser (VWR 890 International LLC, West Chester, PA, USA) for half an hour. Extracts were then filtered using PTFE 0.45 µm syringe filters (FisherbrandTM) to remove insoluble materials. 891 892 DTT determination: The DTT assay was based on the protocol developed by Cho et al. (Cho et 893 al., 2005) and the semi-automated DTT system has been described in detail in Fang et al. 894 (2015b). In brief, 3.5 mL of the aerosol extract, 1 mL potassium phosphate buffer (Kbuffer, 0.5 mM), and 0.5 mL of DTT solution (1mM) were incubated at 37 °C in a continuously mixed vial 895

896 (400 rpm in a ThermoMixer (Eppendorf North America, Inc., Hauppauge, NY, USA)). At five

897 time intervals (4, 13, 23, 30, and 41 min), a small aliquot (100 μL) of the incubated mixture was

898 drawn to another vial and the DTT oxidation was quenched by adding 1 mL 1% w/v

899 Trichloroacetic acid (TCA). 0.5 mL of a color developing agent DTNB [5,5' dithiobis (2-

900 nitrobenzoic acid), 0.2 mM], with 2 mL of Tris buffer (0.08 M with 4 mM EDTA), was added to

901 form a light absorbing product, which has a high extinction coefficient at 412 nm wavelength.

902 Light absorbance at 412 nm and 700 nm (chosen as a baseline) for each time interval was

903 recorded and used to calculate the consumption rate of DTT over time (nmol min⁻¹). Final DTT

904 activity was reported as DTT consumption rate per unit of air volume (nmol min⁺m⁻³), or per</sup>

905 unit of PM mass (nmol min⁻⁺µg⁻⁺). The automated system was cleaned periodically (typically
906 every 15 days) by flushing at least 3 times with methanol and then 6 times with DI water.
907 Sample selection, various chemical reagent addition and withdrawal, and system self cleaning
908 were all achieved automatically by programmable syringe pumps (see details in section 2.2.2).
909 Final DTT activity is calculated as follows:

910
$$\sigma DTT = -\sigma Abs \times \frac{N_{\sigma}}{Abs_{\sigma}} (Eq. 1),$$

911
$$\frac{DTT = \frac{\sigma DTT_s - \sigma DTT_{tr}}{\frac{V_{ct}}{V_{e}} \times V_{p}} (Eq. 2),$$

912 where σ Abs is the slope of absorbance (412nm – 700nm) versus time; Abs₀ is the initial

913 absorbance calculated from the intercept of linear regression of absorbance versus time; N_{θ} is the

914 initial moles of DTT added in the reaction (500 nmol); σDTT_{*}(σDTT_{*}) is the rate of DTT

915 consumption for sample (blank); V_e and V_a are the extraction volume (15 mL) and actual sample

916 volume added to the reaction vial (3.5 mL, i.e., what was actually analyzed), respectively. $V_{\rm p}$ is

917 the ambient air volume (m³) represented by the sample in the extraction volume (V_e, i.e.,

918 considering the filter integration time and fraction of filter analyzed). DTT represents volume

919 normalized DTT activity, in units of nmol min⁻¹m⁻³.

920 2.2.2. Semi-automated system for measuring AA activity

921 Filter extraction: One punch of the collected Hi-Vol filter (5.07 cm²) was extracted in 30 mL of

922 <u>deionized (DI) water (> 18 M Ω cm⁻¹) in a sterile polypropylene centrifuge tube (VWR)</u>

- 923 International LLC, Suwanee, GA, USA) by sonication using an Ultrasonic Cleanser (VWR
- 924 International LLC, West Chester, PA, USA) for half an hour. For those having activities close to

925	blanks, 15 mL was used instead. Extracts were then filtered using PTFE 0.45 µm syringe filters
926	(Fisherbrand TM) to remove insoluble material larger than 0.45 μ m. Although OH may form
927	during sonication (Miljevic et al., 2014), it appears to have little effect on our ROS measurement
928	since we compared the water-soluble ROS activities from the same sample that had been
929	extracted by shaking for 3 hours vs sonication and found no significant differences (average ratio
930	and standard deviation is 1.08 ± 0.20 , $n = 7$.
931	Sample preparation protocol for the AA measurement is the same as that for the DTT assay,
932	except that one punch of the filter (5.07 cm ² each) was extracted in 30 mL of DI water, (for those
933	having activities close to blanks, 15 mL of extraction DI was used instead).
934	AA determination: The method in this study was based on an ascorbate-only model (Mudway et
935	al., 2005; Ayres et al., 2008) which that is a simplified and alternative high throughput approach
936	to a synthetic respiratory tract lining fluid model (RTLF) containing ascorbate, urate, and
937	reduced glutathione (GSH) (Zielinski et al., 1999; Mudway et al., 2004). The semi-automated
938	system for measuring AA activities was adapted from the automated DTT system discussed in
939	section 2.2.1 method protocol is shown in Figure 1 and system setup can be found in the Figure.
940	(Figs. S2 and S3S2 in the Supplement). The method involves two steps.
941	The first step is an aerosol background measurement (Fig. <u>S21</u>). In order to control for the
942	contribution of absorbance of particles themselves at 265 nm wavelength, an AA-free control
943	was measured and subtracted from the sample absorbance readings. 2.4 mL aerosol extracts and
944	0.3 mL 0.5 mM Kbuffer were loaded into a reaction vial (Fig. S3) (sterile polypropylene
945	centrifuge tube, VWR International LLC, Suwanee, GA, USA) using a programmable syringe
946	pump (A) with a 5 mL syringe (Kloehn, Inc., Las Vegas, NV, USA) (Fig. S3). Following mixing,

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947	90 μL of the mixture was transferred to an intermediate vial using Pump B with a 250 μL
948	syringe, and diluted to 3 mL. Pump A then withdrew the diluted mixture from the intermediate
949	vial and pushed it through a Liquid Wave-guide Capillary Cell (LWCC-M-100; World Precision
950	Instruments, Inc., FL, USA) with an optical path length of 100 mm. The waveguide was coupled
951	to an online spectrophotometer, which included a UV-VIS light source (Ocean Optics DT-Mini-
952	2, Ocean Optics, Inc., Dunedin, FL, USA), and a multi-wavelength light detector (USB4000
953	Miniature Fiber Optic Spectrometer, Ocean Optics, Inc., Dunedin, FL, USA). Aerosol
954	background absorbance at 265 and 700 nm (baseline) were recorded at two-second intervals
955	using data acquisition software (SpectraSuite). For the samples collected in this study,
956	backgrounds due to the aerosol absorption at 265 nm were $<10\%$ of the sample absorbance
957	readings. Prior to the second step, the system performed a self-cleaning by flushing the
958	intermediate vial and the two syringes with DI water three times.
959	The second step is the AA measurement (Fig. <u>821</u>). Following the aerosol background
960	measurement, Pump A discarded a fraction of the sample-Kbuffer mixture and left only 1.8 mL
961	in the reaction vial. 0.2 mL 2 mM AA solution was then loaded to the reaction vial using Pump
962	B. Both the reaction and intermediate vial were continuously shaken at 400 rpm in a
963	ThermoMixer (Eppendorf North America, Inc., Hauppauge, NY, USA), which also maintained
964	the incubation temperature at 37 °C. At five different specified times (7, 15, 24, 32, 40 minutes),
965	a small aliquot (100 $\mu L)$ was transferred to the intermediate vial, diluted to 3 mL, and pushed
966	through the LWCC, generating a total of five data points quantifying the remaining AA
967	concentration. The system then again performed a self-cleaning before analyzing the next
968	sample. A multi-position valve (14-port, VICI® Valco Instrument Co. Inc., USA) was used to
969	select samples for analysis. To ensure the suspension of PM in the extract, each sample was

mixed by pushing 5mL of air through the extract before loading to the reaction vial. The detailedKloehn control program code for all steps can be found in the Supporting Information.

972 Final AA activity is calculated as follows:

973

$$\sigma AA = -\sigma Abs \times \frac{N_0}{Abs_0} (\text{Eq. } \underline{31}),$$
974

$$AAv = \frac{\sigma AA_s - \sigma AA_b}{\frac{Va}{Ve} \times V_p} (\text{Eq. } \underline{42}).$$

975 Following the notation above, σAbs is the slope of absorbance versus time, where the absorbance is the absorbance of each time interval subtracting the corresponding aerosol background 976 absorbance; Abso is the initial absorbance calculated from the intercept of linear regression of 977 absorbance versus time; N₀ is the initial moles of AA added in the reaction vial (400 nmol); 978 $\sigma AA_s(\sigma AA_b)$ is the rate of AA consumption for a sample (blank); V_e and V_a are the extraction 979 980 volume (30 or 15 mL) and sample volume added to the reaction (1.6 mL), respectively. V_p is the ambient air volume (m³) represented by the sample in the extraction volume. AAy represents 981 volume normalized AA activity, in units of nmol min⁻¹ m⁻³. Similarly, here DTTy represents the 982 volume normalized DTT activity. 983

984 2.3. Chemical analysis on PM filters

985

2.3.1. Water-soluble organic carbon and brown carbon

An automated system (details in Fig. <u>\$4\$3</u>) was used to measure water-soluble organic carbon
(WSOC) and brown carbon (BrC) on the water-soluble extracts from the same Hi-<u>vol-Vol</u> filters.
Filter extracts (~6mL, same extraction protocol outlined above), after loading onto a 5 mL
sample loop (Upchurch Scientific, Inc., Oak Harbor, WA), were first passed through a 1 m

Liquid Wave guide Capillary Cell <u>LWCC</u> (LWCC-2100; World Precision Instruments, Inc., FL,
USA), where absorbance at 365 nm wavelength (BrC) was measured using an online
spectrophotometer (Ocean Optics, Inc., Dunedin, FL, USA). The extracts then entered a TOC
analyzer (Sievers Model 900, GE Analytical Instruments, Boulder, CO, USA) for determining
WSOC concentration.

995 **2.3.2. Water-soluble elements**

996	A similar automated system was developed to determine the water-soluble elements, including S
997	(Sulfur), Ca (Calcium), K (Potassium), Fe (Iron), Cu (Copper), Zn (Zinc), Ba (Barium), Pb
998	(Lead), As (Arsenic), Sr (Strontium), Se (Selenium), Br (Bromine), Mn (Manganese), and Ti
999	(Titanium). Details of the method are described in Fang et al. (2015a) and in the Supplement.
1000	Filter preparation protocol is the same as described above in section 2.2.1, except that four
1001	punches were extracted in 15 mL of DI water, and after filtering, 120 μ L of high purity HNO ₃
1002	(OmniTrace® Ultra Nitric Acid, 67 70%, EMD Millipore Corporation, Billerica, MA, USA)
1003	was then added to 6 mL of the extract (resulting pH \approx 0.7) to ensure the suspension of all
1004	dissolved metals. Details of the method are described in Fang et al. (2015a). In brief, 5 mL of
1005	acidified sample was aerosolized using a continuous flow ultrasonic nebulizer (CETAC U5000
1006	AT+, CETAC Technologies, Omaha, NE, USA) and directed through an evaporator at $136 ^\circ C$
1007	followed by a condenser at 3 °C. The dry aerosolized sample was then neutralized by a Kr-85 ion
1008	source (Model 3077A, TSI), mixed with filtered make-up air (final flow rate = 16.7 L min ⁻¹ , Pall
1009	HEPA Capsule), and directed to an online aerosol element analyzer (Xact, Cooper
1010	Environmental) for X-ray fluorescence (XRF) analysis.

1011 **2.3.3.** PM_{2.5} mass

1012	PM _{2.5} mass concentration was measured by a Tapered Element Oscillating Microbalance	
1013	(TEOM) by Atmospheric Research Analysis (ARA, Inc.) at SEARCH sites (JST, YRK, BHM,	
1014	and CTR) and ESL. For the RS and GT sites,- <u>since PM_{2.5} mass were not available</u> , the PM mass	
1015	concentrations were estimated from the sum of chemical components analyzed on the same Hi-	
1016	Vol filters . More details can be found in our other publications (Verma et al., 2014; Fang et al.,	
1017	2015b) (Details in the Supplement)and in associated supporting materials.	
1018	2.4. Source apportionment	
1019	Source apportionment of both DTT and AAv activities were was performed using a Positive	
1020	Matrix Factorization (PMF) model (EPA PMF 5.0 software) (Paatero and Tapper, 1994) and a	
1021	Chemical Mass Balance model (version 8.2) with ensemble-averaged source impact profiles	
1022	(CMB-E) (Balachandran et al., 2012). PMF is a commonly used source apportionment approach	
1023	that does not require source profiles as CMB-E, whereas CMB-E has better performance and	
1024	lower relative uncertainties as compared to the PMF method (Balachandran et al., 2012). Source	
1025	contributions to DTTv_activity-using PMF and CMB-E are discussed in our other publications	
1026	(Bates et al., 2015; Verma et al., 2014). A PMF analysis on the water-soluble elements (S, K, Ca,	
1027	Ti, Mn, Fe, Cu, Zn, As, Se, Br, Sr, Ba, and Pb) and WSOC from JST, GT, and RS sites has been	
1028	reported in Fang et al. (2015a). AAv was simply added to the data sets to generate the AAv	
1029	results shown here.	
1030	For the PMF analysis, water soluble AA activity was added to the data sets The PMF results on	
1031	DTTv in our prior analyses were based on JST and GT sitesused in prior analyses (Fang et al.,	
1032	2015a).	

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1033	<u>In PMF, </u> T <u>t</u> he uncertainties for each species were determined by multiplying the concentration by
1034	overall uncertainties (%), which were obtained by propagating the uncertainties from filter
1035	sampling (<u>assumed to be 5%</u>), extraction (<u>assumed to be 5%</u>), blanks (1 σ of multiple blanks),
1036	calibration (1 σ of slope, for water-soluble elements), collocated measurements (for water-
1037	soluble elements and AA, Fig. <u>85S4</u>), and analytical uncertainties. The analytical uncertainties
1038	were obtained by analyzing the same sample/standards multiple times; for example, a composite
1039	of extracts from 11 samples for water-soluble elements (coefficient of variation, $CV = 2-16$ %);
1040	9,10-phenanthrenequinone for AA ($CV = 13\%$), and sucrose standard solutions for WSOC (CV
1041	= 10%). Missing data were replaced by species medians with 400% uncertainty, and values
1042	below LOD were assigned as half of LOD values with uncertainties of 5/6 the concentration
1043	(Polissar et al., 1998). Uncertainty from collocated measurements was calculated as the relative
1044	uncertainty of the slope (1 σ /slope), which was based on an orthogonal regression.
1045	An ensemble-trained source apportionment approach (Balachandran et al., 2012) (CMB-E) was
1046	also used to construct the source impacts of $PM_{2.5}$ -on DTT and $AAv_{\underline{v}}$ -activities-based on $PM_{2.5}$
1047	species (sulfate, nitrate, ammonium, OC, EC, and total metals)and DTT/AAv activity
1048	measured during SCAPE sampling periods (2012-2013). The source profiles cover a range of
1049	sources, including light-duty gasoline vehicles (LDGV), heavy-duty diesel vehicles (HDDV),
1050	ambient sulfate (AMSULF), ambient nitrate (AMNITR), ammonium bisulfate (AMBSLF), not
1051	otherwise apportioned organic carbon (OTHER_OC), dust, biomass burning (BURN), coal fired
1052	power plants (CFPP), cement (CEM), and cooking. An ensemble average was calculated for each
1053	source category using ten different runs developed from four individual source apportionment
1054	methods-(Balachandran et al., 2012). Since the filters for the source impact profiles were
1055	collected from midnight to midnight, while the filters collected for DTT and AAy measurement
	concered from manific to manificate while the metric concered for DTT and Thit, measurement

were collected from noon to 11 a.m. next day, the sources identified were linearly interpolated using a fixed ratio. For example, two consecutive filters (filter 1 and 2) were-collected from midnight to midnight, we would use 12/24* filter 1 + 11/24* filter 2 would be used to produce the estimated DTT or AAv_activities to compared with actual measured data.

1060 **2.5. Epidemiological assessment**

1061

2.5.1. Backcast-estimates of AA and DTT activities

To undertake a time-series epidemiological analysis with sufficient power, retrospective data sets 1062 1063 of daily AAy and DTT levels from 1 August 1998 to 31 December 2009 at an Atlanta site representative of the urban airshed air quality are needed. Previous epidemiological studies by 1064 the study team, assessing Atlanta air quality and emergency department (ED) visits, have used 1065 data from the SEARCH JST site, the anchor site for our DTT and AAy measurements. To 1066 generate daily estimates of retrospective DTT and AAv activities at JST, first a linear model was 1067 used to estimate the contribution of various sources to our observed AAv and DTT activities 1068 1069 measured at JST (Bates et al., 2015). This was done through separate linear regressions for AAy1070 and DTT activities on a per air volume basis, with the ensemble-predicted sources as independent variables. In previous work (Balachandran et al., 2012), a source times-series from 1071 1072 August 1998 to December 2009 was generated for JST using the same CMB-E model with the 1073 same independent variables measured at JST. The DTT and AAy regressions were was then 1074 applied to this time series to construct a time series of estimated DTT and AAy activities for the 1075 epidemiology study time period, during which direct measurements of AAy or DTT-were not 1076 available. In order to test the sensitivity of epidemiologic results to different backcast models,

1077 two other models are generated for AAv (discussed in section 3.2.1). Identical methods were
 1078 applied to DTTv to obtain three different models for comparisons with those from AAv.

2.5.2. Epidemiological analyses

1079

Epidemiological time-series analysis (Strickland et al., 2010; Winquist et al., 2015) was 1080 1081 employed to assess associations of retrospective $DTT_{\underline{v}}$ and $AA\underline{v}$ activities with health effects as reflected in ED visits. Relationships between The study team has published extensively on 1082 1083 previous work analyzing ED visits data from Atlanta area hospitals in relation to and typical 1084 ambient air quality characteristics, as well as the impact of exposure misclassification and other 1085 factors, have been extensively studies (e.g., see Strickland et al., 2015; Darrow et al., 2014; Strickland et al., 2014; Wingquist et al., 2014; -Goldman et al., 2012; Pachon et al., 2012; 1086 1087 Strickland et al., 2011; Strickland et al., 2010; Sarnat et al., 2010; Sarnat et al., 2008; Tolbert et 1088 al., 2007; Metzger et al., 2004; Peel et al., 2005). For the present analysis, we apply our these 1089 previously reported epidemiologic modeling approach to the backcast-estimates of AAy and 1090 DTTv-activities, in order to assess associations of these newly developed air quality descriptors with selected outcomes in the ED visits data collected from hospitals serving the five-county 1091 1092 metropolitan Atlanta area during 1998-2009. The health outcomes investigated in the current 1093 analysis are daily visits for respiratory diseases, including pneumonia (n=145,610 total visits for study period), chronic obstructive pulmonary disease (n=49,251), and asthma/wheeze 1094 1095 (n=263,665), and cardiovascular diseases, including ischemic heart disease (n=73,477) and congestive heart failure (CHF) (n=70,587). The air quality was modeled as a three-day moving 1096 1097 average ("lag 0-2", the moving average of estimated pollutant level for that day, the previous day, and the day before). Poisson generalized linear regression was performed; to control for 1098 1099 temporal trends and meteorological variables, models included cubic splines with monthly knots

1100	for time, linear, quadratic and cubic terms for mean daily dew point (lag 0-2), maximum daily	
1101	temperature (lag 0), and minimum daily temperature (lag 1-2), indicators of hospital contribution	
1102	time periods, season of year, day of week and holiday, and interaction terms between season and	
1103	maximum temperature, and between season and day of week. These covariates were chosen	
1104	based on prior studies (Strickland et al., 2010; Winquist et al., 2015) which identified important	
1105	confounders to the relationship between daily ambient pollution levels and ED visits. Risk ratios	
1106	(the relative risks of ED visit associated with an increase of one interquartile range of the	
1107	exposure metric) and 95% confidence intervals (CI) were used to describe the observed health	
1108	associations. Risk ratios with confidence intervals above 1 are indicative of statistically	
1109	significant positive associations. International classification of disease codes used to define the	
1110	health outcomes can be found in the supplement <u>Supplement</u> .	
1111	3. Results and discussion Discussion	
1111 1112 1113	 Results and <u>discussionDiscussion</u> Comparing DTT and AA based on measurement during 2012-2013<u>AAv for</u> measurements during 2012-2013 and comparisons to DTTv 	
1111 1112 1113 1114	 Results and <u>discussionDiscussion</u> Comparing DTT and AA based on measurement during 2012-2013AAv for measurements during 2012-2013 and comparisons to DTTv Spatio-temporal distribution 	
1111 1112 1113 1114 1115	 3. Results and <u>discussionDiscussion</u> 3.1. Comparing DTT and AA based on measurement during 2012-2013<u>AAv for</u> measurements during 2012-2013 and comparisons to DTTv 3.1.1. Spatio-temporal distribution Monthly average water-soluble AA<u>v</u> and DTT activities at various sampling sites are given in 	
1111 1112 1113 1114 1115 1116	 3. Results and discussionDiscussion 3.1. Comparing DTT and AA based on measurement during 2012-2013AAv for measurements during 2012-2013 and comparisons to DTTv 3.1.1. Spatio-temporal distribution Monthly average water-soluble AAv and DTT activities at various sampling sites are given in Fig. 1-2_(a), and (b), respectively. There were differences in the seasonal and spatial patterns 	
1111 1112 1113 1114 1115 1116 1117	 3. Results and discussionDiscussion 3.1. Comparing DTT and AA based on measurement during 2012-2013AAv for measurements during 2012-2013 and comparisons to DTTv 3.1.1. Spatio-temporal distribution Monthly average water-soluble AAv and DTT activities at various sampling sites are given in Fig. 1-2_(a)and (b), respectively. There were differences in the seasonal and spatial patterns between AA and DTT activities in the Southeastern US. Fig. 1 shows that <u>AAv was</u> 	
1111 1112 1113 1114 1115 1116 1117 1118	 3. Results and discussionDiscussion 3.1. Comparing DTT and AA based on measurement during 2012-2013AAv for measurements during 2012-2013 and comparisons to DTTv 3.1.1. Spatio-temporal distribution Monthly average water-soluble AAy and DTT activities at various sampling sites are given in Fig. +2_(a) and (b), respectively. There were differences in the seasonal and spatial patterns between AA and DTT activities in the Southeastern US. Fig. 1 shows that AAv was heterogeneously distributed, indicated by the significant variability between sites. Highest AAv 	
1111 1112 1113 1114 1115 1116 1117 1118 1119	 3. Results and discussionDiscussion 3.1. Comparing DTT and AA based on measurement during 2012-2013AAv for measurements during 2012-2013 and comparisons to DTTv 3.1.1. Spatio-temporal distribution Monthly average water-soluble AAy and DTT activities at various sampling sites are given in Fig. +2 (a) and (b), respectively. There were differences in the seasonal and spatial patterns between AA and DTT activities in the Southeastern US. Fig. 1 shows that AAv was heterogeneously distributed, indicated by the significant variability between sites. Highest AAv were found at the roadside site (RS) and lowest at rural sites. For example, the ratio of average 	
1111 1112 1113 1114 1115 1116 1117 1118 1119 1120	 3. Results and discussionDiscussion 3.1. Comparing DTT and AA based on measurement during 2012-2013AAv for measurements during 2012-2013 and comparisons to DTTv 3.1.1. Spatio-temporal distribution Monthly average water-soluble AAv and DTT activities at various sampling sites are given in Fig. +2 (a), and (b), respectively. There were differences in the seasonal and spatial patterns between AA and DTT activities in the Southeastern US. Fig. 1 shows that AAv was heterogeneously distributed, indicated by the significant variability between sites. Highest AAv were found at the roadside site (RS) and lowest at rural sites. For example, the ratio of average AAv at RS to its paired Atlanta urban JST site, was 1.2 in fall and 1.6 in winter [(2.7 when 	
1111 1112 1113 1114 1115 1116 1117 1118 1119 1120 1121	 3. Results and discussionDiscussion 3.1. Comparing DTT and AA based on measurement during 2012-2013AAv for measurements during 2012-2013 and comparisons to DTTv 3.1.1. Spatio-temporal distribution Monthly average water-soluble AAy and DTT activities at various sampling sites are given in Fig. +2 (a)_ and (b), respectively. There were differences in the seasonal and spatial patterns between AA and DTT activities in the Southeastern US. Fig. 1 shows that AAv was heterogeneously distributed, indicated by the significant variability between sites. Highest AAv were found at the roadside site (RS) and lowest at rural sites. For example, the ratio of average AAv at RS to its paired Atlanta urban JST site, was 1.2 in fall and 1.6 in winter [(2.7 when including the four high data points in Fig. 2 (a)] and RS to near-road GT was 1.7 in fall 2013. 	

1123	YRK/JST ratio was 0.7 in summer and 0.4 in winter, respectively. An exception is that BHM
1124	(urban) and CTR (rural) had similar AAv (BHM _{avg} = 0.75 \pm 0.45 nmol min ⁻¹ m ⁻³ and CTR _{avg} =
1125	0.78 ± 0.31 nmol min ⁻¹ m ⁻³). Comparing AAv at different sites, BHM and the other urban site,
1126	ESL (average AAv= 0.98 ± 0.63 nmol min ⁻¹ m ⁻³), had lower AAv relative to the Atlanta urban
1127	sites (average of JST and GT in summer = 2.5 ± 1.0 nmol min ⁻¹ m ⁻³). The higher AAv near
1128	traffic sources has also been found in other studies (Janssen et al., 2014; Strak et al., 2012;
1129	Janssen et al., 2015). In contrast, Figure 2(b) shows that DTTy activity was largely spatially
1130	uniform, differences between paired sites is much less than those for AAvexcept the JST-YRK
1131	pair in winter (December) (DTT at JST > DTT at YRK) (also discussed in previous publications
1132	(Verma et al., 2014; Fang et al., 2015b)), while AA was heterogeneously distributed, with the
1133	highest levels near highly trafficked highways and lowest at rural sites. For example, the ratio of
1134	averaged AA activity at RS to its paired Atlanta urban JST site, was 1.2 in fall and 1.6 in winter
1135	{(2.7 when including the four high data points in Fig. 1(a)] and RS to near road GT was 1.7 in
1136	fall 2013. AA was generally lower in rural compared to urban environments (ratio of YRK/JST
1137	is 0.7 in summer and 0.4 in winter, respectively). An exception is that BHM (urban) and CTR
1138	(rural) had similar AA levels (BHM _{avg} = 0.75 ± 0.45 nmol min ⁻¹ m ⁻³ and CTR _{avg} = 0.78 ± 0.31
1139	mmol min ⁻¹ -m ⁻³). BHM and the other urban site, ESL (average AA= 0.98 ± 0.63 nmol min ⁻¹ -m ⁻³),
1140	had lower AA activities relative to the Atlanta urban sites (average of JST and GT in summer $=$
1141	2.5 ± 1.0 nmol min ⁻¹ m ⁻³). The higher AA activities near traffic sources have also been found in
1142	other studies (Janssen et al., 2014; Strak et al., 2012; Janssen et al., 2015).
1143	The seasonal distribution can be examined from the Atlanta sites, JST, GT, RS, and YRK. The
1144	AA activity was higher in summer/fall compared to winter; the ratio of summer or fall to winter

1145 was 1.8, 2.9, 1.0, and 3.1 (average ratio is 2.2 ± 0.9) for JST, YRK, RS, and GT, respectively. In

1146	contrast, JST As discussed in our previous work (Verma et al., 2014; Fang et al., 2015b), there
1147	was high heterogeneity in DTT levels across seasonsDTTv had an opposite seasonal trend, at
1148	JST with the highest level in December (winter), while there was no significant seasonal
1149	variation observed at YRK, GT and the RS site (also seen in Fig. 1). In contrast, the AA activity
1150	was higher in summer/fall compared to winter; the ratio of summer or fall to winter was 1.8, 2.9,
1151	1.0, and 3.1 (average ratio is 2.2 ± 0.9) for JST, YRK, RS, and GT, respectively. These results
1152	indicate that there are differences in the sources for water-soluble AAv and DTTv, with traffic
1153	emissions a more significant source for AAv. Correlation analysis with specific aerosol
1154	components provides further insights. These results indicate that water soluble DTT and AA
1155	activities of PM2.5-were influenced by different aerosol components from differing sources that
1156	varied with season.
1157	3.1.2. Correlations with chemical components
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1157 1158	3.1.2. Correlations with chemical components <u>As an initial step to identifyTo further identify the</u> major sources for <u>DTT and AAv and compare</u>
1157 1158 1159	3.1.2. Correlations with chemical components <u>As an initial step to identifyTo further identify the</u> major sources for <u>DTT and AAv and compare</u> <u>to DTTv</u> -activities, a correlation analysis was performed between the assays and <u>the following</u>
1157 1158 1159 1160	3.1.2. Correlations with chemical components As an initial step to identifyTo further identify the major sources for DTT and AAv and compare to DTTv_activities, a correlation analysis was performed between the assays and the following selected chemical components, including: BrC (an indicator of incomplete combustion, i.e., a
1157 1158 1159 1160 1161	3.1.2. Correlations with chemical components As an initial step to identifyTo further identify the major sources for DTT and AAv and compare to DTTv_activities, a correlation analysis was performed between the assays and the following selected chemical components, including; BrC (an indicator of incomplete combustion, i.e., a biomass burning-indicator), WSOC and S (secondary processes), Ca (mineral dust), and selected
1157 1158 1159 1160 1161 1162	3.1.2. Correlations with chemical components As an initial step to identifyTo further identify the major sources for DTT and AAy and compare to DTTy_activities, a correlation analysis was performed between the assays and the following selected chemical components, including; BrC (an indicator of incomplete combustion, i.e., a biomass burning-indicator), WSOC and S (secondary processes), Ca (mineral dust), and selected transition metals (Cu, Fe, Mn, and Zn) that have been related to adverse health outcomes
1157 1158 1159 1160 1161 1162 1163	3.1.2. Correlations with chemical components As an initial step to identifyTo further identify the major sources for DTT and AAy and compare to DTTy_activities, a correlation analysis was performed between the assays and the following selected chemical components, including; BrC (an indicator of incomplete combustion, i.e., a biomass burning-indicator), WSOC and S (secondary processes), Ca (mineral dust), and selected transition metals (Cu, Fe, Mn, and Zn) that have been related to adverse health outcomes (Cheung et al., 2012; Kam et al., 2011; Shen and Anastasio, 2011; Cheung et al., 2010; Akhtar et
1157 1158 1159 1160 1161 1162 1163 1164	3.1.2. Correlations with chemical components As an initial step to identifyTo further identify the major sources for DTT and AAy and compare to DTTy_activities, a correlation analysis was performed between the assays and the following selected chemical components, including; BrC (an indicator of incomplete combustion, i.e., a biomass burning-indicator), WSOC and S (secondary processes), Ca (mineral dust), and selected transition metals (Cu, Fe, Mn, and Zn) that have been related to adverse health outcomes (Cheung et al., 2012; Kam et al., 2011; Shen and Anastasio, 2011; Cheung et al., 2010; Akhtar et al., 2010; Landreman et al., 2008; Zhang et al., 2008; Kodavanti et al., 2005). Correlation
1157 1158 1159 1160 1161 1162 1163 1164 1165	3.1.2. Correlations with chemical components As an initial step to identifyTo further identify the major sources for DTT and AAy and compare to DTTy_activities, a correlation analysis was performed between the assays and the following selected chemical components, including; BrC (an indicator of incomplete combustion, i.e., a biomass burning-indicator), WSOC and S (secondary processes), Ca (mineral dust), and selected transition metals (Cu, Fe, Mn, and Zn) that have been related to adverse health outcomes (Cheung et al., 2012; Kam et al., 2011; Shen and Anastasio, 2011; Cheung et al., 2010; Akhtar et al., 2010; Landreman et al., 2008; Zhang et al., 2008; Kodavanti et al., 2005). Correlation coefficients based on linear regressions between <u>AAy DTT-</u> or <u>DTTy_AA activity</u> -and chemical
1157 1158 1159 1160 1161 1162 1163 1164 1165 1166	3.1.2. Correlations with chemical components As an initial step to identifyTo further identify the major sources for DTT and AAy and compare to DTTy_activities, a correlation analysis was performed between the assays and the following selected chemical components, including; BrC (an indicator of incomplete combustion, i.e., a biomass burning-indicator), WSOC and S (secondary processes), Ca (mineral dust), and selected transition metals (Cu, Fe, Mn, and Zn) that have been related to adverse health outcomes (Cheung et al., 2012; Kam et al., 2011; Shen and Anastasio, 2011; Cheung et al., 2010; Akhtar et al., 2010; Landreman et al., 2008; Zhang et al., 2008; Kodavanti et al., 2005). Correlation coefficients based on linear regressions between <u>AAy DTT-or DTTy_AA activity</u> -and chemical species (Pearson's r) are shown graphically in Fig. <u>23</u> . A detailed matrix showing the
1157 1158 1159 1160 1161 1162 1163 1164 1165 1166 1167	3.1.2. Correlations with chemical components As an initial step to identifyTo further identify the major sources for DTT and AAv and compare to DTTv_activities, a correlation analysis was performed between the assays and the following selected chemical components, including; BrC (an indicator of incomplete combustion, i.e., a biomass burning-indicator), WSOC and S (secondary processes), Ca (mineral dust), and selected transition metals (Cu, Fe, Mn, and Zn) that have been related to adverse health outcomes (Cheung et al., 2012; Kam et al., 2011; Shen and Anastasio, 2011; Cheung et al., 2010; Akhtar et al., 2010; Landreman et al., 2008; Zhang et al., 2008; Kodavanti et al., 2005). Correlation coefficients based on linear regressions between <u>AAv DTT-</u> or DTTv <u>AA activity</u> and chemical species (Pearson's r) are shown graphically in Fig. <u>23</u> . A detailed matrix showing the correlations at individual sites is given in Table S2. To simplify Figure <u>23</u> , JST and GT were

1169	our previous work for many PM species, such as ECC (Verma et al., 2014), WSOC (Verma et
1170	al., 2014), and water-soluble elements (Fang et al., 2015a: Verma et al., 2014). As shown in Fig.
1171	23, AAv was almost exclusively correlated with water-soluble Cu. The r value ranged from 0.70-
1172	0.94 for most sites/seasons except RS in fall 2012, JST/GT in winter, ESL in summer, and GT in
1173	fall 2013. High correlations between AAv and Cu are consistent with other studies (Janssen et
1174	al., 2014; Künzli et al., 2006), although the correlation coefficients (r) in our work (0.70-0.94)
1175	were higher (0.60-0.74 in other studies), possibly because we used water-soluble Cu and the
1176	other studies used elemental (total) concentrations. Strak et al (2012) also reports a higher r value
1177	between AAv and water-soluble Cu (r = 0.82) than that between AAv and total Cu (r = 0.76)
1178	from the same sample set.
1179	Compared to AAv, DTTv is more broadly correlated with aerosol species: high correlations were
1180	<u>observed</u> DTT activity was correlated with S (r = 0.7 at JST/GT, 0.74 at YRK, 0.66 at BHM ₆₆ -
1181	0.74) and WSOC (r = 0.71 at JST/GT, 0.74 at YRK, $71 0.77$ at CTR) in summer, which
1182	diminished in fall ($r = 0.14-0.66$ for S and $r = 0.20-0.65$ for WSOC) and was weaker in winter
1183	(r<0.4) for S. Instead, higher correlations were found with BrC ($r = 0.78-0.88$) and WSOC ($r = 0.78-0.88$)
1184	0.60-0.84) in winter. The decreasing correlation between DTTv and S going from summer to
1185	winter suggests the important role of secondary processing in summer (Verma et al., 2009a;
1186	McWhinney et al., 2013) and suggesting the important role of secondary processing (Verma et
1187	al., 2009a; McWhinney et al., 2013). These correlations diminished in fall ($r = 0.14$ -0.66 for S
1188	and $r = 0.20$ 0.65 for WSOC) and weak correlations were observed in winter (r<0.4) for S.
1189	Instead, higher correlations were found between DTT and BrC ($r = 0.78 \cdot 0.88$) and WSOC ($r =$
1190	0.60-0.84) in winter. This suggests anthe increasing contribution of biomass burning to winter
1191	DTTv-activities (Verma et al., 2009b); AAv did not show similar trends, i.e., AAv did not

1192	correlate with S (r = -0.12-0.60) and low r values were observed with K in winter (r = $0.07-0.19$,
1193	one exception was JST in December $r = 0.7$), suggesting incomplete combustion (e.g., biomass
1194	burning) was not a significant source for AAv. Whereas AAv was nearly exclusively correlated
1195	with Cu, DTTv was correlated with various metals, including Ca, Mn, Fe, Cu, and Zn. In Fig. 2,
1196	counting the number of sampling sites at which r values between AAv and various metals were
1197	larger than 0.65 (i.e. black solid bars), we observed eight times for Cu and once for Mn ($r =$
1198	0.82). Whereas for DTTv (see the striped bars in Fig.3), we found twice ($r= 0.67$ and 0.77) for
1199	<u>Ca, 3 times for Mn (r = 0.65-0.75), 6 times for Fe (r = 0.68-0.90), once for Cu (r = 0.68), and 3</u>
1200	times for Zn ($r = 0.70-0.82$). DTT activities were also correlated with various metals, including
1201	Ca (r = 0.67–0.77), Mn (r = 0.65–0.75), Fe (r = 0.65–0.74), Cu (r = 0.68), and Zn (r = 0.76–0.82),
1202	but tThere were, however, no apparent seasonal patterns for the correlations between these
1203	water-soluble metals DTT and DTTv these water soluble metals since they were related to
1204	mineral dust (68% of Ca, 45% of Mn, and 26% of Fe) and vehicle brake/tire wear emissions
1205	(51% of Cu, 45% of Zn, 32% of Fe, and 17% of Mn) , as reported by Fang et al. (2015a) .
1206	The comparison of AAv and DTTv's correlation with PM2.5 mass is noteworthy. In contrast, AA
1207	has a much narrower sensitivity to aerosol species than DTT in that it was only correlated with
1208	water soluble Cu (r = 0.70 0.94 for most sites/seasons except JST in Feb, GT in March, and
1209	GT/RS in Sept 2013) (Fig. 2) in the Southeastern US. High correlations between AA and Cu are
1210	eonsistent with other studies (Janssen et al., 2014; Künzli et al., 2006), although the correlation
1211	coefficients (r) in our work (0.70-0.94) were higher (0.60-0.74 in other studies), possibly because
1212	we used water soluble Cu and the other studies used elemental (total) concentrations. Strak et al
1213	(2012) also reports a higher r value between AA and water soluble Cu ($r = 0.82$) than that
1214	between AA and total Cu ($r = 0.76$) from the same sample set.

1215	$DTT\underline{v}$ -activities (per volume) were was fairly well correlated with PM _{2.5} mass (r = 0.49-0.86, Fig.	
1216	32), although the slope of DTT over PM _{2.5} mass differed by season likely due to different	
1217	contributing source mixtures (Fang et al., 2015b).), whereas -In contrast, AAv did not correlate	
1218	as well with PM _{2.5} -mass (r $<0.58 = -0.17$ to 0.59), as also found in Künzli et al. (2006) (r = 0.34).	
1219	$PM_{2.5}$ mass has been linked with adverse health endpoints in many epidemiological studies	
1220	(Laden et al., 2000; Pope et al., 2002; Pope et al., 2004; Metzger et al., 2004; Sarnat et al., 2008),	
1221	thus the <u>lack of correlation</u> of $\frac{DTT}{AAv}$ with PM _{2.5} mass suggests may be suggestive of a lack	
1222	of linkage between DTTAA-measured oxidative potential and health endpoints (tested in section	
1223	3.2). In summary, AA and DTT assays have differing responses-associations to-with PM	
1224	chemical species, further suggested by a lack of strong correlation between the two assays at all	
1225	sites (r<0.55, Fig. 2), similar to findings from other studies (r<0.65) (Janssen et al., 2014; Yang	
1226	et al., 2014; Janssen et al., 2015). A source apportionment analysis is performed to attempt to	
1227	quantify contributions of various sources to PM _{2.5} AAv.	Formatted: Subscript
1227	quantify contributions of various sources to PM _{2.5} AAv. 3.1.3. Source apportionment	Formatted: Subscript
1227 1228	 <u>quantify contributions of various sources to PM_{2.5} AAv.</u> 3.1.3. Source apportionment 	Formatted: Subscript
1227 1228 1229	 <u>quantify contributions of various sources to PM_{2.5} AAv.</u> 3.1.3. Source apportionment Various source contributions to water-soluble AA<u>v</u> and DTT<u>v activities</u> resolved from PMF and 	Formatted: Subscript
1227 1228 1229 1230	quantify contributions of various sources to PM2.5 AAv. 3.1.3. Source apportionment Various source contributions to water-soluble AAv and DTTv activities resolved from PMF and CMB-E based on measured data during 2012-2013 are shown in Fig.3-4 (a, b) and (c, d),	Formatted: Subscript
1227 1228 1229 1230 1231	guantify contributions of various sources to PM2.5 AAv. 3.1.3. Source apportionment Various source contributions to water-soluble AAv and DTTv activities resolved from PMF and CMB-E based on measured data during 2012-2013 are shown in Fig.34 (a, b) and (c, d), respectively. For water soluble AA activityAAv, the detailed loadings of various species and	Formatted: Subscript
1227 1228 1229 1230 1231 1232	guantify contributions of various sources to PM2.5 AAv. 3.1.3. Source apportionment Various source contributions to water-soluble AAy and DTTy activities resolved from PMF and CMB-E based on measured data during 2012-2013 are shown in Fig.3-4 (a, b) and (c, d), respectively. For water soluble AA activityAAv, the detailed loadings of various species and time series of each factor suggested by PMF can be found in Fig. S6S5.	Formatted: Subscript
1227 1228 1229 1230 1231 1232 1233	quantify contributions of various sources to PM2.5 AAv. 3.1.3. Source apportionment Various source contributions to water-soluble AAy and DTTy activities resolved from PMF and CMB-E based on measured data during 2012-2013 are shown in Fig.3-4 (a, b) and (c, d), respectively. For water-soluble AA activityAAv, the detailed loadings of various species and time series of each factor suggested by PMF can be found in Fig. S6S5. AA Sources: Comparing Fig.3-4 (a) and (b) shows that CMB-E and PMF gave consistent and	Formatted: Subscript
1227 1228 1229 1230 1231 1232 1233 1233	quantify contributions of various sources to PM2.5 AAv. 3.1.3. Source apportionment Various source contributions to water-soluble AAy and DTTy activities resolved from PMF and CMB-E based on measured data during 2012-2013 are shown in Fig.3-4 (a, b) and (c, d), respectively. For water soluble AA activityAAy, the detailed loadings of various species and time series of each factor suggested by PMF can be found in Fig. S6S5. AA Sources: Comparing Fig.3-4 (a) and (b) shows that CMB-E and PMF gave consistent and complementary results for AAy sources. Consistent with the spatial distributions, CMB-E	Formatted: Subscript
1227 1228 1229 1230 1231 1232 1233 1234 1235	quantify contributions of various sources to PM2.5 AAv. 3.1.3. Source apportionment Various source contributions to water-soluble AAy and DTTy activities-resolved from PMF and CMB-E based on measured data during 2012-2013 are shown in Fig.3-4 (a, b) and (c, d), respectively. For water soluble AA activityAAv, the detailed loadings of various species and time series of each factor suggested by PMF can be found in Fig. S6S5. AA Sources: Comparing Fig.3-4 (a) and (b) shows that CMB-E and PMF gave consistent and complementary results for AAy sources. Consistent with the spatial distributions, CMB-E indicated that vehicles comprise almost half of the total source contributions to PM2.5 AAy	Formatted: Subscript
1227 1228 1229 1230 1231 1232 1233 1234 1235 1236	quantify contributions of various sources to PM2.5 AAV. 3.1.3. Source apportionment Various source contributions to water-soluble AAy and DTTy activities resolved from PMF andCMB-E based on measured data during 2012-2013 are shown in Fig.3-4 (a, b) and (c, d),respectively. For water soluble AA activityAAv, the detailed loadings of various species andtime series of each factor suggested by PMF can be found in Fig. S6S5.AA Sources: Comparing Fig.3-4 (a) and (b) shows that CMB-E and PMF gave consistent andcomplementary results for AAy sources. Consistent with the spatial distributions, CMB-Eindicated that vehicles comprise almost half of the total source contributions to PM2.5 AAyactivity with roughly equal contributions from light and heavy duty vehicles. PMF also found	Formatted: Subscript

1238	(44% from brake/tire wear). Both methods also found sources of $AA\underline{v}$ associated with secondary
1239	processes. CMB-E provides more details in that it separated out secondary organic and ambient
1240	sulfate sources. This is consistent with our source apportionment analyses on water-soluble
1241	metals (Fang et al., 2015a), which showed that Cu was mainly associated with secondary
1242	formation and brake/tire wear, consistent with $AA\underline{v}$ being highly correlated with Cu. The 19%
1243	other OC source in Figure 3b is related to un-apportioned OC, which and so-includes mainly
1244	secondary biogenic organic aerosols from biogenic emissions, and possible additional
1245	contributions from other VOC sources. At this point, the role it plays as a source of $AA\underline{v}$ is not
1246	clear. It should be noted that the residual for PMF was -10% and that for CMB-E was 14%,
1247	which means the PMF model over-predicted and CMB-E under-estimated AAv-activity. The
1248	source apportionment analysis is consistent with the spatial distribution, which indicated vehicle
1249	emissions as a main source for AAv activity.
1250	AAy compared to DTTy Sources: Comparisons of the The-source apportionment results on
1250 1251	AAy compared to DTTy Sources: <u>Comparisons of the The</u> -source apportionment results on water-soluble <u>AADTTy activity</u> -using PMF_(Verma et al., 2014)-and CMB-E to a similar
1250 1251 1252	AAy compared to DTTy Sources: <u>Comparisons of the The</u> -source apportionment results on water-soluble <u>AADTTy activity</u> -using PMF(Verma et al., 2014)-and CMB-E to a similar analyses for DTTv is insightfulwere compared to those for AA. PMF source apportionment
1250 1251 1252 1253	AAy compared to DTTy Sources: <u>Comparisons of the The</u> -source apportionment results on water-soluble <u>AADTTy activity</u> using PMF_(Verma et al., 2014) and CMB-E to a similar analyses for DTTv is insightfulwere compared to those for AA. PMF source apportionment analyses [Fig. <u>34</u> (a) and (c)] suggest a common contribution from traffic emissions and
1250 1251 1252 1253 1254	AAy compared to DTTy Sources: Comparisons of the The-source apportionment results on water-soluble <u>AADTTy</u> activity using PMF_(Verma et al., 2014) and CMB-E to a similar analyses for DTTv is insightfulwere compared to those for AA. PMF source apportionment analyses [Fig. <u>34</u> (a) and (c)] suggest a common contribution from traffic emissions and secondary processes to both water-soluble <u>AAv_DTT-</u> and <u>DTTvAA activities</u> , but the
1250 1251 1252 1253 1254 1255	AAy compared to DTTy Sources: Comparisons of the The-source apportionment results on water-soluble <u>AADTTy activity using PMF_(Verma et al., 2014)</u> and CMB-E to a similar analyses for DTTv is insightfulwere compared to those for AA. PMF source apportionment analyses [Fig. <u>34</u> (a) and (c)] suggest a common contribution from traffic emissions and secondary processes to both water-soluble <u>AAv_DTT</u> and <u>DTTvAA activities</u> , but the contributions were stronger for AAy than DTTy. For example, 44% AAy activity was attributed
1250 1251 1252 1253 1254 1255 1256	AAy compared to DTTy Sources: Comparisons of the The-source apportionment results on water-soluble <u>AADTTy activity using PMF_(Verma et al., 2014)</u> and CMB-E to a similar analyses for DTTv is insightfulwere compared to those for AA. PMF source apportionment analyses [Fig. <u>34</u> (a) and (c)] suggest a common contribution from traffic emissions and secondary processes to both water-soluble <u>AAv_DTT</u> and <u>DTTvAA activities</u> , but the contributions were stronger for AAy than DTTv. For example, 44% AAy <u>activity</u> was attributed to vehicles and 56% to secondary processes, compared to 16% and 31% <u>of-for_DTTv_activities</u> ,
1250 1251 1252 1253 1254 1255 1256 1257	AAy compared to DTTy Sources: Comparisons of the The-source apportionment results on water-soluble <u>AADTTy activity using PMF_(Verma et al., 2014)</u> and CMB-E to a similar analyses for DTTv is insightfulwere compared to those for AA. PMF source apportionment analyses [Fig. <u>34</u> (a) and (c)] suggest a common contribution from traffic emissions and secondary processes to both water-soluble <u>AAv_DTT</u> and <u>DTTvAA activities</u> , but the contributions were stronger for AAy than DTTy. For example, 44% AAy <u>activity</u> was attributed to vehicles and 56% to secondary processes, compared to 16% and 31% <u>of-for</u> DTTy-activities, respectively. <u>Higher fractional contributions of these two sources for AAv is because unlike</u>
1250 1251 1252 1253 1254 1255 1256 1257 1258	AAy compared to DTTy Sources: Comparisons of the The-source apportionment results on water-soluble AADTTy activity-using PMF_(Verma et al., 2014)-and CMB-E to a similar analyses for DTTv is insightfulwere compared to those for AA. PMF source apportionment analyses [Fig. 34(a) and (c)] suggest a common contribution from traffic emissions and secondary processes to both water-soluble AAv_DTT- and DTTvAA activities, but the contributions were stronger for AAy than DTTy. For example, 44% AAy activity was attributed to vehicles and 56% to secondary processes, compared to 16% and 31% of-for_DTTy-activities, respectively. Higher fractional contributions of these two sources for AAv is because unlike DTTv, biomass burning does not contribute to AAv (1%), whereas it makes a large contribution
1250 1251 1252 1253 1254 1255 1256 1257 1258 1259	AAy compared to DTTy Sources: Comparisons of the The-source apportionment results on water-soluble AADTTy activity-using PMF_(Verma et al., 2014) and CMB-E to a similar analyses for DTTv is insightful were compared to those for AA. PMF source apportionment analyses [Fig. 34(a) and (c)] suggest a common contribution from traffic emissions and secondary processes to both water-soluble AAv_DTT and DTTvAA activities, but the contributions were stronger for AAy than DTTy. For example, 44% AAy activity-was attributed to vehicles and 56% to secondary processes, compared to 16% and 31% of-for DTTy-activities, respectively. Higher fractional contributions of these two sources for AAv is because unlike DTTv, biomass burning does not contribute to AAv (1%), whereas it makes a large contribution to the overall study DTTv (35%). CMB-E also found no contribution of biomass burning to
1250 1251 1252 1253 1254 1255 1256 1257 1258 1259 1260	AAy compared to DTTy Sources: Comparisons of the The-source apportionment results on water-soluble AADTTy activity using PMF_(Verma et al., 2014) and CMB-E to a similar analyses for DTTv is insightfulwere compared to those for AA. PMF source apportionment analyses [Fig. 34(a) and (c)] suggest a common contribution from traffic emissions and secondary processes to both water-soluble AAv_DTT-and DTTvAA activities, but the contributions were stronger for AAy than DTTy. For example, 44% AAy activity was attributed to vehicles and 56% to secondary processes, compared to 16% and 31% of for DTTy-activities, respectively. Higher fractional contributions of these two sources for AAv is because unlike DTTv, biomass burning does not contribute to AAv (1%), whereas it makes a large contribution to the overall study DTTv (35%). CMB-E also found no contribution of biomass burning to AAv, but identified a fractional contribution from biomass burning [36% BURN in Fig. 4(d)] to

1261	DTTv similar to PMF. This is because a large fraction of the DTT activities (35%) were related to
1262	biomass burning while only 1% of AA activities were attributed to biomass burning. CMB-E
1263	also found similar contribution from biomass burning [36% BURN in Fig. 3(d)] to DTT
1264	activities. CMB-E points to ammonium sulfate (AMSULF) as a source for both AAv and DTTv.
1265	Neither AA nor DTT assay responds to pure ammonium sulfate, meaning that ammonium sulfate
1266	is an indicator of some source or process. It may be a marker for atmospheric processed or aged
1267	aerosols. For example, both assays respond to water-soluble transition metals, and a significant
1268	fraction these metals, when emitted, are not water-soluble [solubilities of Zn is ~50%, Cu and
1269	Mn 10-40%, Fe < 10% (Birmili et al., 2006; Espinosa et al., 2002)]. Mobilization by acidic
1270	aerosols can increase the soluble fraction, which requires a low aerosol pH and time, both can be
1271	linked to sulfate aerosol. For example, at pH < 2 , 1-2% of mineral dust Fe is mobilized within 3-
1272	5 days (Meskhidze et al., 2003).
1273	It is worth noting that both assays appear to be linked to emissions from traffic, but the actual
1274	sources from traffic differ. AAv was almost exclusively associated with mechanically generated
1275	aerosols (i.e. brake/tire wear), whereas for water-soluble DTTv, traffic emissions included both
1276	metals and organic aerosol species, i.e., from mechanically generated (brake/tire wear) and
1277	combustion (tail pipe emissions). Finally, correlations to specific aerosol species and source
1278	apportionment analysis can be confounded by co-variability with other unmeasured components
1279	or processes, as demonstrated by the associations with ammonium sulfate, or nonlinear responses
1280	of these assays to specific components (Charrier et al., 2015). However, the major sources
1281	identified for both AAv and DTTv, and the contrasts between their sources, is consistent with the
1282	season trends and spatial distributions observed and discussed above for each assay, indicating
1283	that the source apportionment analysis is robust. The common contribution of ammonium sulfate

1284	(AMSULF) to DTT and AA may represent different sources, i.e., organic aerosol oxidation or
1285	metal mobilization by formation of an aqueous particle with secondary acids. Note that for
1286	water-soluble DTT, traffic emissions included both metals and organic aerosol species, that is
1287	from mechanically generated (brake/tire wear) and combustion (tail pipe emissions),
1288	respectively, whereas AA was almost exclusively associated with mechanically generated
1289	aerosols (i.e. brake/tire wear).
1290 1291	3.2. AA <u>v</u> and DTT activities compared to association with health endpoints and application to epidemiological models contrasts to DTTv
1292	3.2.1. Backcast-estimates of AA <u>v</u> and DTT activities-using Source Impacts
1293	Although over roughly 1 year of AAv of DTT and AA activities were generated for the central
l 1294	JST site in Atlanta, longer data sets are generally needed for a time series epidemiological study.
1295	To generate these data, <u>a</u> multiple linear regression s were was used to estimate AAv and DTT
1296	activities from the CMB-E identified sources. We follow the same approach as that used for
1297	DTTv (Bates et al., 2015). Water-soluble AAv and DTT activities (nmol min ⁻¹ m ⁻³) measured
1298	between 2012 and 2013 at JST were regressed against all CMB-E sources. Insignificant sources
1299	(p of F-statistic of coefficient>0.05, Table S3) and the significant sources with negative
1300	coefficients were removed. The latter occurred for BURN (biomass burning) and AMNITR
1301	(ammonium nitrate) in the AA regression, likely due to their opposite seasonal trends to the
1302	measured AAv activity. These two sources also did not contribute to AAv [see Fig. 4(a)]Biomass
1303	burning contributed only 1% and ammonium nitrate did not contribute to AA activity [see Fig.
1304	$\frac{3(a)}{a}$. The final regressions for AA <u>v</u> and DTT activities are is:

 $AAAA_{\nu}^{e} = 0.079 + 0.19 LDGV + 0.23 HDDV + 0.063 AMSULF + 0.075 OTHER_OC (Eq. <u>53</u>)$

1306	For direct comparison with DTTv, we used the same criteria for including various sources in the	
1307	DTTv regression model, with the result:	
1308	<u>ــــــــــــــــــــــــــــــــــــ</u>	Formatted: Font: 12 pt
1309	$\frac{DTT}{V}DTT_{v}^{e} = 0.067 + 0.11 LDGV + 0.045 HDDV + 0.02 AMSULF + 0.069 BURN \text{ (Eq. 64)}$	
1310	(Note, the DTTv regression is different from that in Bates et al., (2015) in that AMSULF (ammonium	
1311	sulfate) was included in this model)	
1312	AA_v^e where DTT and DTT $_v^eAA$ are the estimated ROS activities of PM _{2.5} (nmol min ⁻¹ m ⁻³), which	
1313	are related to the following sources ($\mu g m^{-3}$): light-duty gasoline vehicles (LDGV), heavy-duty	
1314	diesel vehicles (HDDV), ammonium sulfate (AMSULF), biomass burning (BURN), and other	
1315	organic carbon (OTHER_OC). The coefficients in the equations represent the intrinsic activities	
1316	(nmol min ⁻¹ μ g ⁻¹) of the sources, a measure of the strength of the source on a per PM _{2.5} mass	
1317	basis for water-soluble $AA\underline{v}$ or $DTT\underline{v}$ -activities. Interestingly, for both assays, the traffic sources	
1318	(LDGV+HDDV) has the highest ROS intrinsic activity, while secondary sources or biomass	
1319	burning have relatively lower ROS intrinsic activities. The high intrinsic activity in the traffic	
1320	sources might be attributed to metals that have much higher intrinsic ROS activities The high	
1321	DTT intrinsic activity in the traffic source might be attributed to metals that have much higher	
1322	intrinsic DTT activities (Charrier and Anastasio, 2012; Verma et al., 2015). The much higher	Field Code Changed
1323	coefficients of LDGV and HDDV in the AA_v^e regression than those in DTT_v^e highlight the larger	Field Code Changed Field Code Changed
1324	role of metals from these sources contributing to the overall AAe. AAlthough biomass burning	
1325	has a lower DTT intrinsic activity compared to the other sources in the DTTv model, it was the	
1326	largest contributor to $DTT_v^e DTT$ activity due to the strength of this source over the measurement	
1327	period (e.g. large magnitude of BURN). (Bates et al., 2015). For AA, vehicles and secondary	
-		

1328	formation had the largest contribution. The regression positive intercepts (0.079 and 0.067 for
1329	AA and DTT regression, respectively) indicate some some levels of discrepancy between the
1330	model and measurementunidentified source for AAv and DTTv. The ability of the models to
1331	predict AA_v^e and DTT_v^e is given by the correlation between the model and measurements (AA_v^e vs
1332	<u>AAv and DTTv</u> , and vs DTTv). The r r-values (are 0.60 and 0.68 for AAvA and DTTv DTT,
1333	respectively) between measured and predicted activities represent the uncertainties, indicating
1334	the models can only account for about 40% of the observed variabilityRegression coefficients,
1335	p-values, and the correlation coefficients (r) r values are summarized between measured and
1336	estimated data are in Table S3.
1337	To test the sensitivity of the epidemiological results to other predictive models, two other
1338	regressions were used to predict <u>AAv DTT</u> and <u>DTTv AA</u> : 1) all sources included, 2) only
1339	significant sources with positive coefficients (above), but with AMSULF removed (summarized
1340	in Table S3). The latter was done because sulfate has substantially decreased over the last decade
1341	due to emission reductions in the southeastern US (Hidy et al., 2014; Hand et al., 2012), which
1342	may have unknown effects on AA_v^e and DTT_v^e . These models and various statistics are also
1343	summarized in Table S3. For both assays the models with all sources included had highest
1344	correlations coefficients between model and observed activities (r ~ 0.7 , or model explains ~ 50%
1345	of the variability).
 1346	3.2.2. Health associations from time-series epidemiological models

Backcast AA^e_vAA and DTT^e_vDTT activities were next generated for the study period
corresponding to the health analysis(ED) data. The various regression models (including from
EEq.5-3 and 6-4) were used to generate daily retrospective estimates of AAv and DTTv activities
at the JST site for the period of 1998-2009, -(Bates et al., 2015), based on existing source impacts

1351	generated in a previous study for the same site in the Southeastern US (Balachandran et al.,	
1352	2012). Estimated water-soluble AA_v^e DTT and DTT_v^e AA activities were run separately in	
1353	epidemiological models of ED visits for selected outcomes (section 2.5.2). The risk ratios for	
1354	AA_v^e DTT and DTT A activities for asthma/wheeze and congestive heart failure are presented	
1355	in Figure 4-5 (data given in Table S4). The other health outcomes (chronic obstructive	
1356	pulmonary disease, pneumonia, and ischemic heart disease) did not show significant associations	
1357	with AA_v^e estimated DTT or DTT $_v^e$ AA activity (results given in Table S4 not presented).	
1358	For asthma/wheeze and congestive heart failure, although the risk ratios for an increase of an	
1359	interquartile range for $\frac{\text{DTT-AA}_{v}^{\text{e}}}{\text{estimated from the model (Eq.6)}}$ were above 1 [(Fig. 41.005 and	
1360	1.003 for Asthma/wheeze and CHF, respectively, Fig. 5(a))], and the 95 % confidence intervals	
1361	exclude crossed 1 (0.994-1.015 and 0.986-1.020 for Asthma/wheeze and congestive heart failure,	
1362	<u>respectively</u>), indicating a <u>non-statistically significant positive</u> association between AA_v^e	
1363	estimated DTT-and the ED visits for these health outcomes. In contrast, neither-both of the health	
1364	outcomes showed statistically significant associations with the DTT_v^eAA estimates. Similar The	
1365	same results differences were found for estimates based on the two other regressions [(see Figure	
1366	$\frac{S75(b)}{S75(b)}$ and (c)], suggesting that the null relationship of AA _v and positive association of DTT _v ^e	
1367	with these health outcomes are to some extent robust, despite the high uncertainties from the	
1368	<u>back-cast models.</u> A possible cause for the differences in $\frac{DTT-AA_v}{v}$ and $\frac{AA-DTT_v}{v}$ health	
1369	associations is, at least -differences in uncertainties in the estimated historical ROS activities; the	
1370	model did not observed AA activities as well as DTT, (observed vs predicted r values 0.60 and	
1371	0.68 for AA and DTT, respectively). Ffor this study region, the more narrow selectivity of the	
1372	AA assay to specific aerosol components (i.e., mostly sensitive to Cu). The may result in AA	

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1373	assay may not capturing capture the integrated overall oxidative potential of all the various PM	
1374	ROS activity of all the various PM components as well as the DTT assay	
1375	PM-induced oxidative stress in the exacerbation of asthma (Li et al., 2003) and the	
1376	cardiovascular system (Donaldson et al., 2001) has been proposed. Linkages seen here between	
1377	water-soluble ROS activity and morbidity due to asthma/wheeze and congestive heart failure are	
1378	consistent with these studies and the contrast with the AAv results suggest the importance of	
1379	organic components and transition metals from biomass burning and vehicular emissions in the	
1380	Southeastern US, and support aerosol particle oxidative potential as a mechanism contributing to	
1381	these PM-induced adverse health effects. Although some studies have shown that ROS plays a	
1382	key role in COPD (O'Donnell et al., 2006), IHF (Lakshmi et al., 2009; Giordano, 2005), and	_
1383	Pneumonia (Kuwano et al., 2003), we did not observed a significantly positive association of the	
1384	tested ROS assays with these health outcomes. Finally, although this work shows a contrast	
1385	between these two assays and association with health endpoints, Janssen et al. (2015) found	
1386	significant associations between both assays and nasal and airway inflammation based on a	
1387	different approach (panel study, n=31).	
1388	4. Summary	
1389	Approximately 500 PM _{2.5} high-volume filter samples collected in the Southeastern US were	
1390	analyzed for aerosol oxidative potential using the DTT and AA assays. The AA activities	

- 1391 reported are from the same filters for which water-soluble DTT activities had already been
- 1392 <u>determined.</u> We found that <u>water-soluble AA activity on a per air volume basis (AAv) was</u>
- 1393 <u>highest near roadways and lowest at rural sites. AAv was higher in summer/fall than winter.</u>
- 1394 <u>These results are in contrast to DTTv, which was more spatially uniform and had an opposite</u>

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1395	seasonal trend at the urban Atlanta site (higher in winter than summer/fall).water soluble DTT
1396	activity (nmol min ⁻¹ -m ⁻³) was fairly spatially uniform and higher in winter than in summer/fall at
1397	the urban Atlanta site, whereas AAv was most consistently correlated with water-soluble metals
1398	(especially water-soluble Cu), whereas water soluble AA activity had highest levels near
1399	roadways, and an opposite seasonal trend (higher in summer/fall than winter). DTT \underline{v} activity was
1400	correlated with organic speciesand_water-soluble metals (Fe, Cu, Zn, Mn, and Ca), and also
1401	$PM_{2.5}$ mass, whereas AA activity was only correlated with water-soluble metals (especially
1402	water soluble Cu). A Source apportionment analysis suggest indicated that traffic emissions
1403	secondary processes and secondary processes traffic emissions were strong contributors to both
1404	AAv_DTT-and DTTv_AA activities-in urban Atlanta, GA, with traffic emissions the most
1405	important. For AAv only road dust was responsible, in contrast to both combustion emissions
1406	and road dust contributing to the DTTv from this source. Biomass burning was did not contribute
1407	to AAv, but was a large-substantial source for DTTv-activity, but insignificant for AAconsistent
1408	with AAv being mainly associated with transition metals. These source apportionment results are
1409	also consistent with observed seasonal trends and spatial distributions, for both assays. Multiple
1410	linear regressions suggest that mobile sources have the highest ROS intrinsic activities (i.e.,
1411	activities per PM _{2.5} -mass from that source) for both assays. Time-series large population
1412	epidemiological analyses using backcast backcast-estimates of AAv DTT-and AA-DTTv from a
1413	number of linear models activities based on 10-year historical source impacts suggest that DTT
1414	AAv activity was not associated linked with any emergency department (ED) visits for both
1415	asthma/wheeze and congestive heart failureall tested health outcomes at 95% confidence
1416	intervals. DTTv was associated with ED visits for both asthma/wheeze and congestive heart
1417	failure, for all the linear models tested. Neither AAv nor DTTv was associated with chronic
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1418	obstructive pulmonary disease (COPD), Ischemic heart failure (IHD) or pneumonia at a
1419	statistically significant level. Based on the wide-ranging comparisons between these assays, we
1420	conclude that, , while AA activity was not linked to any health outcomes. Ffor the region
1421	investigated in this study, the DTT assay was a more comprehensive multi-pollutant indicator of
1422	ROS (or PM2.5 oxidative potential) indicator than the AA assay. This can be useful when deciding
1423	on what assay to utilize to address the goal of a specific studymaking DTT is a potentially a
1424	more valuable parameter to include in future PM health-related studies because of its broader
1425	sensitivity to aerosol components associated with oxidative potential. Finally, T the ability to
1426	readily measure both <u>PM_{2.5}</u> AA and DTT with automated systems enables activities can support
1427	further larger scale studies of involving direct measurements of PM oxidative potential. These
1428	types of future studies are needed to test these measures and linkages to health outcomes to assess
1429	if these our health findings based on backcast-estimated AA and DTT levels are robust and
1430	applicable tofound in other regions.
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1440 References

- 1441 Akhtar, U. S., McWhinney, R. D., Rastogi, N., Abbatt, J. P. D., Evans, G. J., and Scott, J. A.: Cytotoxic
- 1442 and proinflammatory effects of ambient and source-related particulate matter (PM) in relation to the 1443 production of reactive oxygen species (ROS) and cytokine adsorption by particles, Inhalation Toxicology,
- 1444 22, 37-47, 10.3109/08958378.2010.518377, 2010.
- 1445 Aust, A., Ball, J., Hu, A., Lighty, J., Smith, K., Straccia, A., Veranth, J., and Young, W.: Particle
- 1446 characteristics responsible for effects on human lung epithelial cells, Res Rep Health Eff Inst, 1 65,1447 2002.
- 1448 Ayres, J. G., Borm, P., Cassee, F. R., Castranova, V., Donaldson, K., Ghio, A., Harrison, R. M., Hider,
- R., Kelly, F., and Kooter, I. M.: Evaluating the toxicity of airborne particulate matter and nanoparticles by
 measuring oxidative stress potential-a workshop report and consensus statement, Inhalation toxicology,
 20, 75-99, 2008.
- Balachandran, S., Pachon, J. E., Hu, Y., Lee, D., Mulholland, J. A., and Russell, A. G.: Ensemble-trained
- source apportionment of fine particulate matter and method uncertainty analysis, Atmospheric Environment, 61, 387-394, http://dx.doi.org/10.1016/j.atmoseny.2012.07.031, 2012.
- 1455 Bates, J.T., Weber, R. J., Abrams, J., Verma, V., Fang, T., Klein, M, Strickland, M. J., Sarnat, S. E.,
- 1456 Chang, H. H., Mulholland, J. A., Tolbert, P. E., Russell, A. G.: Reactive Oxygen Species Generation
- Linked to Soures of Atmospheric Particulate Matter and Cardiorespiratory Effects, Environmental
 Science & Technology, in review.
- 1459 Birmili, W., Allen, A. G., Bary, F., and Harrison, R. M.: Trace Metal Concentrations and Water Solubility
- in Size-Fractionated Atmospheric Particles and Influence of Road Traffic, Environ. Sci. Techno., 40,
 144-1153, doi:10.1021/es0486925, 2006.
- Brunekreef, B., Janssen, N. A., de Hartog, J., Harssema, H., Knape, M., and van Vliet, P.: Air pollution
 from truck traffic and lung function in children living near motorways, Epidemiology, 8, 298-303,
- 1464 10.1097/00001648-199705000-00012, 1997.
- 1465 Burchiel, S. W., Lauer, F. T., Dunaway, S. L., Zawadzki, J., McDonald, J. D., and Reed, M. D.:
- 1466 Hardwood smoke alters murine splenic T cell responses to mitogens following a 6-month whole body
- 1467 inhalation exposure, Toxicology and Applied Pharmacology, 202, 229-236,
- 1468 http://dx.doi.org/10.1016/j.taap.2004.06.024, 2005.

- 1469 Charrier, J. G., and Anastasio, C.: On dithiothreitol (DTT) as a measure of oxidative potential for ambient
- 1470 particles: evidence for the importance of soluble transition metals, Atmos. Chem. Phys., 12, 9321-9333,
- 1471 10.5194/acp-12-9321-2012, 2012.
- 1472 Cheung, K., Shafer, M. M., Schauer, J. J., and Sioutas, C.: Diurnal trends in oxidative potential of coarse
- particulate matter in the Los Angeles basin and their relation to sources and chemical composition,
 Environmental Science & Technology, 46, 3779-3787, 10.1021/es204211v, 2012.
- 1474 Environmental Science & Technology, 46, 3779-3787, 10.1021/es204211v, 2012.
 1475 Cheung K L, Ntziachristos L, Tzamkiozis T, Schauer J L, Samaras Z, Moore
- Cheung, K. L., Ntziachristos, L., Tzamkiozis, T., Schauer, J. J., Samaras, Z., Moore, K. F., and Sioutas,
 C.: Emissions of particulate trace elements, metals and organic species from gasoline, diesel, and
- biodiesel passenger vehicles and their relation to oxidative potential, Aerosol Science and Technology,
- 1478 44, 500-513, 10.1080/02786821003758294, 2010.
- 1479 Cho, A. K., Sioutas, C., Miguel, A. H., Kumagai, Y., Schmitz, D. A., Singh, M., Eiguren-Fernandez, A.,
- 1480 and Froines, J. R.: Redox activity of airborne particulate matter at different sites in the Los Angeles
- 1481 Basin, Environmental research, 99, 40-47, 2005.
- 1482 Chung, M. Y., Lazaro, R. A., Lim, D., Jackson, J., Lyon, J., Rendulic, D., and Hasson, A. S.: Aerosol-
- borne quinones and reactive oxygen species generation by particulate matter extracts, Environmental
 Science & Technology, 40, 4880-4886, 10.1021/es0515957, 2006.
- 1485 Darrow, L. A., Klein, M., Flanders, W. D., Mulholland, J. A., Tolbert, P. E., and Strickland, M. J.: Air
- Pollution and Acute Respiratory Infections Among Children 0–4 Years of Age: An 18-Year Time-Series
 Study, American Journal of Epidemiology, 180, 968-977, 10.1093/aje/kwu234, 2014.
- 1488 DiStefano, E., Eiguren-Fernandez, A., Delfino, R. J., Sioutas, C., Froines, J. R., and Cho, A. K.:
- Determination of metal-based hydroxyl radical generating capacity of ambient and diesel exhaust
 particles, Inhal Toxicol, 21, 731-738, 10.1080/08958370802491433, 2009.
- Donaldson, K., Stone, V., Seaton, A., and MacNee, W.: Ambient particle inhalation and the
- 1491 Donaudson, K., Stone, Y., Scatol, A., and Wactvee, W.: Amblent particle initiation and the
 1492 cardiovascular system: potential mechanisms, Environmental Health Perspectives, 109, 523-527, 2001.
- Donaldson, K., Tran, L., Jimenez, L., Duffin, R., Newby, D., Mills, N., MacNee, W., and Stone, V.:
- Combustion-derived nanoparticles: a review of their toxicology following inhalation exposure, Part Fibre Toxicol, 2, 10, 10.1186/1743-8977-2-10, 2005.
- Espinosa, A. J. F., Rodríguez, M. T., de la Rosa, F. J. B., and Sánchez, J. C. J.: A chemical speciation of trace metals for fine urban particles, Atmos. Environ., 36, 773–780, doi:10.1016/S1352-2310(01)00534-9, 2002.
- 1499 Fang, T., Guo, H., Verma, V., Peltier, R. E., and Weber, R. J.: PM2.5 water-soluble elements in the
- southeastern United States: automated analytical method development, spatiotemporal distributions,
 source apportionment, and implications for heath studies, Atmos. Chem. Phys. Discuss., 15, 17189-
- 1502 17227, 10.5194/acpd-15-17189-2015, 2015a.
- 1503 Fang, T., Verma, V., Guo, H., King, L. E., Edgerton, E. S., and Weber, R. J.: A semi-automated system
- for quantifying the oxidative potential of ambient particles in aqueous extracts using the dithiothreitol
 (DTT) assay: results from the Southeastern Center for Air Pollution and Epidemiology (SCAPE), Atmos.
- 1506 Meas. Tech., 8, 471-482, 10.5194/amt-8-471-2015, 2015b.
- 1507 Gasser, M., Riediker, M., Mueller, L., Perrenoud, A., Blank, F., Gehr, P., and Rothen-Rutishauser, B.:
- Toxic effects of brake wear particles on epithelial lung cells in vitro, Particle and Fibre Toxicology, 6, 30,
 10.1186/1743-8977-6-30, 2009.
- Giordano, F. J.: Oxygen, oxidative stress, hypoxia, and heart failure, Journal of Clinical Investigation,
 115, 500-508, 10.1172/JCI200524408, 2005.
- 1512 Goldman, G. T., Mulholland, J. A., Russell, A. G., Gass, K., Strickland, M. J., and Tolbert, P. E.:
- 1513 Characterization of ambient air pollution measurement error in a time-series health study using a
- 1514 geostatistical simulation approach, Atmospheric Environment, 57, 101-108,
- 1515 10.1016/j.atmosenv.2012.04.045, 2012.
- 1516 Hand, J. L., Schichtel, B. A., Malm, W. C., and Pitchford, M. L.: Particulate sulfate ion concentration and
- 1517 SO2 emission trends in the United States from the early 1990s through 2010, Atmos. Chem. Phys., 12,
- **1518** <u>10353-10365, 10.5194/acp-12-10353-2012, 2012.</u>

- Hansen, D. A., Edgerton, E. S., Hartsell, B. E., Jansen, J. J., Kandasamy, N., Hidy, G. M., and Blanchard, 1519
- 1520 C. L.: The Southeastern aerosol research and characterization study: Part 1-Overview, Journal of the Air
- 1521 & Waste Management Association, 53, 1460-1471, 10.1080/10473289.2003.10466318, 2003.
- Harkema, J. R., Keeler, G., Wagner, J., Morishita, M., Timm, E., Hotchkiss, J., Marsik, F., Dvonch, T., 1522
- 1523 Kaminski, N., and Barr, E .: Effects of concentrated ambient particles on normal and hypersecretory
- 1524 airways in rats, Res Rep Health Eff Inst, 120, 1-68, 2004.
- Harrison, R. M., and Yin, J.: Particulate matter in the atmosphere: which particle properties are important 1525 1526 for its effects on health?, Science of The Total Environment, 249, 85-101,
- 1527 http://dx.doi.org/10.1016/S0048-9697(99)00513-6, 2000.
- 1528 Hidy, G. M., Blanchard, C. L., Baumann, K., Edgerton, E., Tanenbaum, S., Shaw, S., Knipping, E.,
- 1529 Tombach, I., Jansen, J., and Walters, J.: Chemical climatology of the southeastern United States, 1999-1530 2013, Atmos. Chem. Phys., 14, 11893-11914, 10.5194/acp-14-11893-2014, 2014.
- 1531 Janssen, N. A. H., Yang, A., Strak, M., Steenhof, M., Hellack, B., Gerlofs-Nijland, M. E., Kuhlbusch, T.,
- Kelly, F., Harrison, R., Brunekreef, B., Hoek, G., and Cassee, F.: Oxidative potential of particulate matter 1532
- collected at sites with different source characteristics, Science of The Total Environment, 472, 572-581, 1533 1534 http://dx.doi.org/10.1016/j.scitotenv.2013.11.099, 2014.
- 1535
- Janssen, N. A. H., Strak, M., Yang, A., Hellack, B., Kelly, F. J., Kuhlbusch, T. A. J., Harrison, R. M., 1536 Brunekreef, B., Cassee, F. R., Steenhof, M., and Hoek, G.: Associations between three specific a-cellular
- 1537 measures of the oxidative potential of particulate matter and markers of acute airway and nasal
- 1538 inflammation in healthy volunteers, Occupational and environmental medicine, 72, 49-56, 2015.
- 1539 Jung, H., Guo, B., Anastasio, C., and Kennedy, I. M.: Quantitative measurements of the generation of
- 1540 hydroxyl radicals by soot particles in a surrogate lung fluid, Atmos. Environ., 40, 1043-1052, 2006.
- 1541 Kam, W., Ning, Z., Shafer, M. M., Schauer, J. J., and Sioutas, C.: Chemical characterization and redox
- 1542 potential of coarse and fine particulate matter (PM) in underground and ground-level rail systems of the 1543 Los Angeles metro, Environmental Science & Technology, 45, 6769-6776, 10.1021/es201195e, 2011.
- 1544 Kelly, F. J., Mudway, I., Blomberg, A., Frew, A., and Sandström, T.: Altered lung antioxidant status in
- 1545 patients with mild asthma, The Lancet, 354, 482-483, http://dx.doi.org/10.1016/S0140-6736(99)01812-7, 1999. 1546
- 1547 Kleinman, M. T., Sioutas, C., Froines, J. R., Fanning, E., Hamade, A., Mendez, L., Meacher, D., and
- 1548 Oldham, M.: Inhalation of concentrated ambient particulate matter near a heavily trafficked road
- 1549 stimulates antigen-induced airway responses in mice, Inhalation toxicology, 19, 117-126,
- 1550 10.1080/08958370701495345, 2007.
- Kodavanti, U. P., Schladweiler, M. C., Ledbetter, A. D., McGee, J. K., Walsh, L., Gilmour, P. S., 1551
- 1552 Highfill, J. W., Davies, D., Pinkerton, K. E., Richards, J. H., Crissman, K., Andrews, D., and Costa, D.
- 1553 L.: Consistent pulmonary and systemic responses from inhalation of fine concentrated ambient particles:
- 1554 roles of rat strains used and physicochemical properties, Environmental Health Perspectives, 113, 1561-1555 1568, 10.1289/ehp.7868, 2005.
- 1556 Kumagai, Y., Koide, S., Taguchi, K., Endo, A., Nakai, Y., Yoshikawa, T., and Shimojo, N.: Oxidation of
- proximal protein sulfhydryls by phenanthraquinone, a component of diesel exhaust particles, Chemical 1557
- Research in Toxicology, 15, 483-489, 10.1021/tx0100993, 2002. 1558
- Künzli, N., Mudway, I. S., Götschi, T., Shi, T., Kelly, F. J., Cook, S., Burney, P., Forsberg, B., 1559
- 1560 Gauderman, J. W., Hazenkamp, M. E., Heinrich, J., Jarvis, D., Norbäck, D., Payo-Losa, F., Poli, A.,
- 1561 Sunyer, J., and Borm, P. J. A.: Comparison of Oxidative Properties, Light Absorbance, and Total and 1562 Elemental Mass Concentration of Ambient PM(2.5) Collected at 20 European Sites, Environmental
- 1563 Health Perspectives, 114, 684-690, 10.1289/ehp.8584, 2006.
- 1564 Kuwano, K., Nakashima, N., Inoshima, I., Hagimoto, N., Fujita, M., Yoshimi, M., Maeyama, T.,
- 1565 Hamada, N., Watanabe, K., and Hara, N.: Oxidative stress in lung epithelial cells from patients with
- idiopathic interstitial pneumonias, European Respiratory Journal, 21, 232-240, 1566
- 1567 10.1183/09031936.03.00063203, 2003.

- Laden, F., Neas, L. M., Dockery, D. W., and Schwartz, J.: Association of fine particulate matter from
 different sources with daily mortality in six U.S. cities, Environmental Health Perspectives, 108, 941-947,
- different sources with daily mortality in six U.S. cities10.2307/3435052, 2000.
- 1571 Lakshmi, S. V., Padmaja, G., Kuppusamy, P., and Kutala, V. K.: Oxidative stress in cardiovascular
- 1572 <u>disease</u>, Indian journal of biochemistry & biophysics, 46, 421-440, 2009.
- 1573 Landreman, A. P., Shafer, M. M., Hemming, J. C., Hannigan, M. P., and Schauer, J. J.: A macrophage-
- based method for the assessment of the Reactive Oxygen Species (ROS) activity of atmospheric
- 1575 particulate matter (PM) and application to routine (daily-24 h) aerosol monitoring studies, Aerosol
- 1576 Science and Technology, 42, 946-957, 10.1080/02786820802363819, 2008.
- 1577 Li, N., Hao, M., Phalen, R. F., Hinds, W. C., and Nel, A. E.: Particulate air pollutants and asthma: A
- paradigm for the role of oxidative stress in PM-induced adverse health effects, Clinical Immunology, 109,
 250-265, <u>http://dx.doi.org/10.1016/j.clim.2003.08.006</u>, 2003.
- Lin, P., and Yu, J. Z.: Generation of reactive oxygen species mediated by Humic-like Substances in
 atmospheric aerosols, Environmental Science & Technology, 45, 10362-10368, 10.1021/es2028229,
 2011.
- 1583 Lundstedt, S., White, P. A., Lemieux, C. L., Lynes, K. D., Lambert, I. B., Öberg, L., Haglund, P., and
- Tysklind, M.: Sources, fate, and toxic hazards of oxygenated polycyclic aromatic hydrocarbons (PAHs) at
 PAH- contaminated sites, AMBIO: A Journal of the Human Environment, 36, 475-485, 10.1579/0044-
- 1586 7447(2007)36[475:sfatho]2.0.co;2, 2007.
- McWhinney, R. D., Zhou, S., and Abbatt, J. P. D.: Naphthalene SOA: redox activity and naphthoquinone
 gas-particle partitioning, Atmos. Chem. Phys., 13, 9731-9744, 10.5194/acp-13-9731-2013, 2013.
- Meskhidze, N., Chameides, W. L., Nenes, A., and Chen, G.: Iron mobilization in mineral dust: Can
 anthropogenic SO2 emissions affect ocean productivity?, Geophysical Research Letters, 30 (21).
- doi:10.1029/2003GL018035, 2003.
- 1592 Metzger, K. B., Tolbert, P. E., Klein, M., Peel, J. L., Flanders, W. D., Todd, K., Mulholland, J. A., Ryan,
- 1593 P. B., and Frumkin, H.: Ambient air pollution and cardiovascular emergency department visits,
- 1594 Epidemiology, 15, 46-56, 10.1097/01.EDE.0000101748.28283.97, 2004.
- 1595 Miljevic, B., F. Hedayat, S. Stevanovic, K. E. Fairfull-Smith, S. E. Bottle, and Z. D. Ristovski (2014), To
- 1596 Sonicate or Not to Sonicate PM Filters: Reactive Oxygen Species Generation Upon Ultrasonic Irradiation,
- 1597 <u>Aerosol Science and Technology</u>, 48(12), DOI:10.1080/02786826.2014.981330, 1276-1284.
- Mudway, I., Duggan, S., Venkataraman, C., Habib, G., Kelly, F., and Grigg, J.: Combustion of dried
 animal dung as biofuel results in the generation of highly redox active fine particulates, Particle and Fibre
 Toxicology, 2, 6, 10.1186/1743-8977-2-6, 2005.
- 1601 Mudway, I. S., Stenfors, N., Duggan, S. T., Roxborough, H., Zielinski, H., Marklund, S. L., Blomberg,
- 1602 A., Frew, A. J., Sandström, T., and Kelly, F. J.: An in vitro and in vivo investigation of the effects of
- diesel exhaust on human airway lining fluid antioxidants, Archives of Biochemistry and Biophysics, 423,
 200-212, <u>http://dx.doi.org/10.1016/j.abb.2003.12.018</u>, 2004.
- 1605 Nel, A.: Air pollution-related illness: effects of particles, Science, 308, 804-806,
- 1606 10.1126/science.1108752, 2005.
- 1607 Nel, A. E., Diaz-Sanchez, D., and Li, N.: The role of particulate pollutants in pulmonary inflammation
 and asthma: evidence for the involvement of organic chemicals and oxidative stress, Current opinion in
- pulmonary medicine, 7, 20-26, 10.1097/00063198-200101000-00004, 2001.
 O'Denraf D. Wilson D. Wilson C. and Dividence in the signature in COPD.
- <u>O'Donnell, R., Breen, D., Wilson, S., and Djukanovic, R.: Inflammatory cells in the airways in COPD,</u>
 <u>Thorax, 61, 448-454, 10.1136/thx.2004.024463, 2006.</u>
- 1612 Paatero, P., and Tapper, U.: Positive matrix factorization: A non-negative factor model with optimal
- 1613 utilization of error estimates of data values, Environmetrics, 5, 111-126, 10.1002/env.3170050203, 1994.
- 1614 Pachon, J. E., Balachandran, S., Hu, Y., Mulholland, J. A., Darrow, L. A., Sarnat, J. A., Tolbert, P. E.,
- and Russell, A. G.: Development of Outcome-based, Multipollutant Mobile Source Indicators, Journal of
 the Air & Waste Management Association, 62, 431-442, 10.1080/10473289.2012.656218, 2012.

Formatted: Normal

- 1617 Peel, J. L., Tolbert, P. E., Klein, M., Metzger, K. B., Flanders, W. D., Todd, K., Mulholland, J. A., Ryan,
- 1618 P. B., and Frumkin, H.: Ambient air pollution and respiratory emergency department visits,
- 1619 Epidemiology, 16, 164-174, 10.1097/01.ede.0000152905.42113.db, 2005.
- 1620 Polissar, A. V., Hopke, P. K., Paatero, P., Malm, W. C., and Sisler, J. F.: Atmospheric aerosol over
- Alaska: 2. Elemental composition and sources, Journal of Geophysical Research: Atmospheres, 103, 19045-19057, 10.1029/98JD01212, 1998.
- 1623 Pope, C. A., Burnett, R. T., Thun, M. J., Calle, E. E., Krewski, D., Ito, K., and Thurston, G. D.: Lung
- 1624 cancer, cardiopulmonary mortality, and long-term exposure to fine particulate air pollution, Journal of the
 1625 American Medical Association, 287, 1132-1141, 10.1001/jama.287.9.1132, 2002.
- 1626 Pope, C. A., Burnett, R. T., Thurston, G. D., Thun, M. J., Calle, E. E., Krewski, D., and Godleski, J. J.:
- 1627 Cardiovascular mortality and long-term exposure to particulate air pollution: epidemiological evidence of 1628 general pathophysiological pathways of disease, Circulation, 109, 71-77,
- 1629 10.1161/01.CIR.0000108927.80044.7F, 2004.
- 1630 Roginsky, V. A., Barsukova, T. K., and Stegmann, H. B.: Kinetics of redox interaction between
- substituted quinones and ascorbate under aerobic conditions, Chem Biol Interact, 121, 177-197, 1999.
- 1632 Samet, J. M., Dominici, F., Curriero, F. C., Coursac, I., and Zeger, S. L.: Fine particulate air pollution and
- 1633 mortality in 20 U.S. cities, 1987–1994, The New England Journal of Medicine, 343, 1742-1749,
- 1634 doi:10.1056/NEJM200012143432401, 2000.
- 1635 Sarnat, J. A., Marmur, A., Klein, M., Kim, E., Russell, A. G., Sarnat, S. E., Mulholland, J. A., Hopke, P.
- K., and Tolbert, P. E.: Fine particle sources and cardiorespiratory morbidity: an application of chemical
 mass balance and factor analytical source-apportionment methods, Environ Health Perspect, 116, 459466, 10.1289/ehp.10873, 2008.
- 1038 400, 10.1289/cmp.10875, 2008.
- 1639 Sarnat, S. E., Klein, M., Sarnat, J. A., Flanders, W. D., Waller, L. A., Mulholland, J. A., Russell, A. G.,
- and Tolbert, P. E.: An examination of exposure measurement error from air pollutant spatial variability in
 time-series studies, Journal of exposure science & environmental epidemiology, 20, 135-146,
- 1642 10.1038/jes.2009.10, 2010.
- 1643 Sauvain, J.-J., Deslarzes, S., and Riediker, M.: Nanoparticle reactivity toward dithiothreitol,
- 1644 Nanotoxicology, 2, 121-129, 10.1080/17435390802245716, 2008.
- Schaumann, F., Borm, P. J. A., Herbrich, A., Knoch, J., Pitz, M., Schins, R. P. F., Luettig, B., Hohlfeld, J.
 M., Heinrich, J., and Krug, N.: Metal-rich ambientparticles (particulate matter2.5) cause airway
- inflammation in healthy subjects, American Journal of Respiratory and Critical Care Medicine, 170, 898 903, 10.1164/rccm.200403-423OC, 2004.
- Shen, H., and Anastasio, C.: Formation of hydroxyl radical from San Joaquin Valley particles extracted in
 a cell-free surrogate lung fluid, Atmospheric chemistry and physics 11, 9671-9682, 10.5194/acp-11-9671 2011, 2011.
- 1652 Shi, T., Schins, R. P., Knaapen, A. M., Kuhlbusch, T., Pitz, M., Heinrich, J., and Borm, P. J.: Hydroxyl
- 1653 radical generation by electron paramagnetic resonance as a new method to monitor ambient particulate
- 1654 matter composition, Journal of environmental monitoring : JEM, 5, 550-556, 10.1039/B303928P, 2003.
- 1655 Strak, M., Janssen, N. A., Godri, K. J., Gosens, I., Mudway, I. S., Cassee, F. R., Lebret, E., Kelly, F. J.,
- 1656 Harrison, R. M., Brunekreef, B., Steenhof, M., and Hoek, G.: Respiratory health effects of airborne
- 1657 particulate matter: the role of particle size, composition, and oxidative potential-the RAPTES project,
- 1658 Environ Health Perspect, 120, 1183-1189, 10.1289/ehp.1104389, 2012.
- 1659 Strickland, M. J., Darrow, L. A., Klein, M., Flanders, W. D., Sarnat, J. A., Waller, L. A., Sarnat, S. E.,
- 1660 Mulholland, J. A., and Tolbert, P. E.: Short-term Associations between Ambient Air Pollutants and
- Pediatric Asthma Emergency Department Visits, American Journal of Respiratory and Critical Care
 Medicine, 182, 307-316, 10.1164/rccm.200908-1201OC, 2010.
- 1663 Strickland, M. J., Darrow, L. A., Mulholland, J. A., Klein, M., Flanders, W. D., Winquist, A., and
- 1664 Tolbert, P. E.: Implications of different approaches for characterizing ambient air pollutant concentrations
- 1665 within the urban airshed for time-series studies and health benefits analyses, Environmental Health, 10,
- 1666 36, 10.1186/1476-069X-10-36, 2011.

- 1667 Strickland, M. J., Klein, M., Flanders, W. D., Chang, H. H., Mulholland, J. A., Tolbert, P. E., and
- 1668 Darrow, L. A.: Modification of the Effect of Ambient Air Pollution on Pediatric Asthma Emergency
- Visits: Susceptible Subpopulations, Epidemiology, 25(6), 843-850, 10.1097/EDE.0000000000170,
 2014.
- 1670 2014.1671 Strickland, M. J.,
- Strickland, M. J., Gass, K. M., Goldman, G. T., and Mulholland, J. A.: Effects of ambient air pollution
 measurement error on health effect estimates in time-series studies: a simulation-based analysis, J Expos
- 1673 Sci Environ Epidemiol, 25, 160-166, 10.1038/jes.2013.16, 2015.
- 1674 Tolbert, P. E., Klein, M., Peel, J. L., Sarnat, S. E., and Sarnat, J. A.: Multipollutant modeling issues in a
- study of ambient air quality and emergency department visits in Atlanta, Journal of Exposure Science and
 Environmental Epidemiology, 17, S29-S35, 10.1038/sj.jes.7500625, 2007.

Valko, M., Morris, H., and Cronin, M. T.: Metals, toxicity and oxidative stress, Current medicinal
 chemistry, 12, 1161-1208, 10.2174/0929867053764635 2005.

- 1679 Vejerano, E. P., Ma, Y., Holder, A. L., Pruden, A., Elankumaran, S., and Marr, L. C.: Toxicity of
- particulate matter from incineration of nanowaste, Environmental Science: Nano, 2, 143-154,
 10.1039/C4EN00182F, 2015.
- Venkatachari, P., Hopke, P. K., Grover, B. D., and Eatough, D. J.: Measurement of particle-bound
 reactive oxygen species in rubidoux aerosols, J. Atmos. Chem., 50, 49–58, 2005.
- Verma, V., Ning, Z., Cho, A. K., Schauer, J. J., Shafer, M. M., and Sioutas, C.: Redox activity of urban
 quasi-ultrafine particles from primary and secondary sources, Atmospheric Environment, 43, 6360-6368,
 10.1016/j.atmosenv.2009.09.019, 2009e.
- 1687 Verma, V., Polidori, A., Schauer, J. J., Shafer, M. M., Cassee, F. R., and Sioutas, C.: Physicochemical 1688 and toxicological profiles of particulate matter in Los Angeles during the October 2007 southern
- 1689 California wildfires, Environmental Science & Technology, 43, 954 960, 10.1021/es8021667, 2009b.
- 1690 Verma, V., Rico-Martinez, R., Kotra, N., King, L. E., Liu, J., Snell, T. W., and Weber, R. J.: Contribution
- of water-soluble and insoluble components and their hydrophobic/hydrophilic subfractions to the reactive
 oxygen species-generating ootential of fine ambient aerosols, Environmental Science & Technology, 46,
- 1693 11384-11392, 10.1021/es302484r, 2012.
- 1694 Verma, V., Fang, T., Guo, H., King, L. E., Bates, J. T., Peltier, R. E., Edgerton, E. S., Russell, A. G., and
- 1695 Weber, R. J.: Reactive oxygen species associated with water-soluble PM2.5 in the southeastern United 1696 States: spatiotemporal trends and source apportionment, Atmos. Chem. Phys., 14, 12915-12930,
- 1697 10.5194/acp-14-12915-2014, 2014.
- Verma, V., Fang, T., Xu, L., Peltier, R. E., Russell, A. G., Ng, N. L., and Weber, R. J.: Organic aerosols
 associated with the generation of Reactive Oxygen Species (ROS) by water-soluble PM2.5,
- 1700 Environmental Science & Technology, 49, 4646-4656, 10.1021/es505577w, 2015.
- 1701 Winquist, A., Kirrane, E., Klein, M., Strickland, M., Darrow, L. A., Sarnat, S. E., Gass, K., Mulholland,
- J., Russell, A., and Tolbert, P.: Joint Effects of Ambient Air Pollutants on Pediatric Asthma Emergency
 Department Visits in Atlanta, 1998–2004, Epidemiology, 25(5), 666-673, doi:
- 1704 10.1097/EDE.00000000000146, 2014.
- 1705 Winquist, A., Schauer, J. J., Turner, J. R., Klein, M., and Sarnat, S. E.: Impact of ambient fine particulate
- 1706 matter carbon measurement methods on observed associations with acute cardiorespiratory morbidity, J
- 1707 Expos Sci Environ Epidemiol, 25, 215-221, 10.1038/jes.2014.55, 2015.
- 1708 Yang, A., Jedynska, A., Hellack, B., Kooter, I., Hoek, G., Brunekreef, B., Kuhlbusch, T. A. J., Cassee, F.
- 1709 R., and Janssen, N. A. H.: Measurement of the oxidative potential of PM2.5 and its constituents: The
- 1710 effect of extraction solvent and filter type, Atmospheric Environment, 83, 35-42,
- 1711 <u>http://dx.doi.org/10.1016/j.atmosenv.2013.10.049</u>, 2014.
- 1712 Zhang, Y., Schauer, J. J., Shafer, M. M., Hannigan, M. P., and Dutton, S. J.: Source apportionment of in
- 1713 vitro Reactive Oxygen Species bioassay activity from atmospheric particulate matter, Environmental
- 1714 Science & Technology, 42, 7502-7509, 10.1021/es800126y, 2008.
- 1715 Zielinski, H., Mudway, I. S., Berube, K. A., Murphy, S., Richards, R., and Kelly, F. J.: Modeling the
- interactions of particulates with epithelial lining fluid antioxidants, The American journal of physiology,277

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1718 1719 1720 1721 1722	Zomer, B., Collé, L., Jedy'nska, A., Pasterkamp, G., Kooter, I., and Bloemen, H.: Chemiluminescent reductive acridinium triggering (CRAT) – mechanism and applications, Anal. Bioanal. Chem., 401, 2945–2954, 2011., L719-726, 1999.
1723	



1728	Figure 12. Monthly average (\pm SD) of PM _{2.5} oxidative potential based on the (a) AA and (b)
1729	DTT assays from the water-soluble extracts from filters collected at three urban (JST, BHM, and
1730	ESL), two rural (YRK and CTR), a near-road (GT), and a road-side (RS) site in the Southeastern
1731	United States.
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Figure 23. Correlation coefficient (Pearson's r) of fine particle water-soluble AA or DTT



1741 A more detailed correlation table is provided in Table S2.



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1743 Figure 34. Contribution of various factors resolved by PMF (a, c), and ensemble (b, d), to the
1744 water-soluble AA (a, b) and DTT (c, d) activities measured during 2012-2013. BURN – biomass
1745 burning; AMSULF – ammonium sulfate; HDDV – heavy-duty diesel vehicles; LDGV – light-
1746 duty gasoline vehicles; OTHER_OC – other organic carbon which secondary organic aerosols
1747 from biogenic emissions, and possible additional contributions from other VOC sources contains
1748 mostly biogenic carbons.
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1756 positive coefficients; (b) all sources; and (c) significant positive sources without AMSULF

<u>(ammonium sulfate).</u> The models were generated from a multiple regression of the measured
 <u>AA activities DTT</u> or <u>DTT-AA activities</u>, on a per volume air bases, with all sources from CMB-

1759 E as independent variables. Risk ratios and associated 95% confidence intervals are presented for

an increase of one interquartile range (IQR) increment of the exposure metric. A risk ratio with

- 1761 95% confidence intervals (CI) for interquartile range above 1 indicates a statistically significant
- 1762 positive association. Risk ratio data <u>and related statistics</u> can be found in Table S4. Results from

1763 other models are shown in Fig S7.