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Maternal Exposure to Particulate Air Pollution and Term Birth Weight: A Multi-Country Evaluation of Effect and Heterogeneity

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BACKGROUND: A growing body of evidence has associated maternal exposure to air pollution with adverse effects on fetal growth; however, the existing literature is inconsistent.

OBJECTIVES: We aimed to quantify the association between maternal exposure to particulate air pollution and term birth weight and low birth weight (LBW) across 14 centers from 9 countries, and to explore the influence of site characteristics and exposure assessment methods on between-center heterogeneity in this association.

METHODS: Using a common analytical protocol, International Collaboration on Air Pollution and Pregnancy Outcomes (ICAPPO) centers generated effect estimates for term LBW and continuous birth weight associated with PM10 and PM2.5 (particulate matter ≤ 10 and 2.5 µm). We used meta-analysis to combine the estimates of effect across centers (~3 million births) and used meta-regression to evaluate the influence of center characteristics and exposure assessment methods on between-center heterogeneity in reported effect estimates.

RESULTS: In random-effects meta-analyses, term LBW was positively associated with a 10-µg/m3 increase in PM10 [odds ratio (OR) = 1.03; 95% CI: 1.01, 1.05] and PM2.5 (OR = 1.10; 95% CI: 1.03, 1.18) exposure during the entire pregnancy, adjusted for maternal socioeconomic status. A 10-µg/m3 increase in PM2.5 exposure was also negatively associated with term birth weight as a continuous outcome (the fully adjusted analysis revealed a 34-g decrease per 10-µg/m3 increase in PM2.5; 95% CI: −13.2, −4.6 g). Meta-regressions revealed that centers with higher median PM2.5 levels and PM2.5:PM10 ratios, and centers that used a temporal exposure assessment (compared with spatiotemporal), tended to report stronger associations.

CONCLUSION: Maternal exposure to particulate pollution was associated with LBW at term across study populations. We detected three site characteristics and aspects of exposure assessment methodology that appeared to contribute to the variation in associations reported by centers.


The developing fetus is known to be susceptible to environmental insults (Stillerman et al. 2008). A growing body of evidence has associated maternal exposure to ambient air pollution with a range of adverse pregnancy outcomes including low birth weight (LBW), intra-uterine growth retardation, preterm birth, stillbirth, and congenital anomalies (Glinianaia et al. 2004; Sapkota et al. 2010; Šrámk et al. 2005; Vrijheid et al. 2011). However, notable inconsistencies among the findings of these studies (Parker and Woodruff 2008; Parker et al. 2011; Woodruff et al. 2009) have hindered the ability of policy makers to incorporate the research evidence into policy. Discrepancies among previous studies may reflect genuine differences in the study settings, may be a consequence of specific biases, or may arise from differences in study designs and exposure assessments (Parker and Woodruff 2008; Woodruff et al. 2009). Study setting characteristics that may contribute to variation in reported associations include the demographic characteristics of the study population, the major sources of pollutants, the size distribution of particulate pollutants [e.g., PM2.5:PM10 ratio (particulate matter ≤ 2.5 and 10 µm)], maternal time–activity patterns, the study period, the degree of confounding by socioeconomic status (SES), and the underlying prevalence of adverse pregnancy outcomes. Relevant study design characteristics include the sources of...
data for fetomaternal characteristics (e.g.,
birth certificates, questionnaires, or hospital
records), inclusion/exclusion criteria, outcome
definitions (e.g., birth weight as a continuous
variable, LBW, or small for gestational age),
and the analysis of potential confounders and/
or effect modifiers. Differences in exposure
assessment include applied methods for assess-
ing the exposure (e.g., proximity-based, moni-
tor-based, or model-based methods), exposure
time windows, exposure contrasts (e.g., spa-
tial, temporal, or spatiotemporal), the avail-
ability of data for specific pollutants, and the
analysis of associations with combinations of
pollutants (Parker and Woodruff 2008).

The International Collaboration on Air Pollution
and Pregnancy Outcomes (ICAPPO) was estab-
lished to investigate the association between maternal exposure to
ambient air pollution and pregnancy out-
comes across multiple centers and to under-
stand how differences in study settings and
methods contribute to variations in findings
(i.e., between-center heterogeneity) (Parker
To achieve this goal, we previously discussed
methodological differences in the published
studies (Woodruff et al. 2009), described the
collaborative centers (Woodruff et al. 2010),
and presented preliminary estimates of effects
for each center (Parker et al. 2011).

The overarching aim of this analysis was
to evaluate the association between maternal
exposure to particulate air pollution and term
birth weight and LBW. Toward this aim, we
combined effect estimates of the individual
ICAPPO centers and assessed the between-
center heterogeneity in these associations.
The application of a common analysis protocol
across ICAPPO centers provided a unique
opportunity to separate the contribution of
the analytical design to between-center hetero-
genesis, and enabled us to assess the impact of
center characteristics and exposure assessment
methods on the variation in reported effect
estimates by each center.

Methods

Overview. Our study was based on estimates of
effects [odds ratio (OR) for LBW and
regression coefficients for birth weight] that
were uniformly generated and reported by
each ICAPPO center according to a com-
mon protocol (Parker et al. 2011). Ambient
levels of PM$_{10}$ and PM$_{2.5}$ were used as indica-
tors of particulate air pollution. Our analysis
was focused principally on the association
between maternal exposure to PM$_{10}$ during
the entire pregnancy and term LBW (birth
weight < 2500 g at 37–42 completed weeks of
gestation) because this was reported by
most ICAPPO centers. We also conducted
additional analyses of data from subsets of
ICAPPO centers that stratified PM$_{10}$–term
LBW analyses by the exposure time window
(i.e., the first, second, and third trimester),
analyzed birth weight as a continuous out-
come variable, and estimated the associa-
tion between maternal exposure to PM$_{2.5}$ and
term LBW. We synthesized the effect estimates across the centers by applying a
meta-analysis framework. The effects of cen-
ter characteristics and exposure assessment
methods on between-center heterogeneity in
effect estimates were explored using a meta-
regression framework. The meta-analyses and
meta-regressions were conducted using the R
statistical package (http://cran.r-project.org/),
libraries meta, rmeta, and metafor.

ICAPPO. Our analysis relied on effect esti-
mates provided by fourteen ICAPPO centers
from nine countries with more than three mil-
lion singleton term births (Table 1). For the
ICAPPO analysis, the centers reanalyzed exist-
ing data sets that had been created to evaluate
the impacts of maternal exposure to air pol-
novation outcomes. The centers relied on outcome data available from rou-
tinely collected administrative records (birth
certificates) or data collected for a specific study
(Bell et al. 2007, 2008; Brauer et al. 2008;
Darrow et al. 2011; Gehring et al. 2011;
Glinianaia et al. 2008; Gouveia et al. 2004; Ha
et al. 2004; Jalaludin et al. 2007; Lepeule
et al. 2010; Mannes et al. 2005; Morello-Frosch
et al. 2010; Pesatori et al. 2008; Rich et al.
2009; Slama et al. 2009; van den Hooven
et al. 2009). More detailed description of the
ICAPPO centers has been previously published
(Parker et al. 2011; Woodruff et al. 2010).

Participating centers were initially asked
to provide information on their study location
and period, available air pollutants, number of
births, prevalence of term LBW, exposure assess-
ment method, and available covari-
data (Parker et al. 2011; Woodruff et al.
2010). Based on this information, ICAPPO
participants developed a common analyti-
cal protocol detailing the inclusion criteria,
outcomes and covariates of interest, statistical
models, and sensitivity analyses. This protocol
also specified a standardized way of reporting
the results (Parker et al. 2011). Each center
was asked to reanalyze its existing data set
according to this protocol. The analyses were
limited to live-born, singleton, term births
with known birth weight, maternal education
(or another measure of SES), dates of birth
and conception, and ambient PM$_{10}$ or PM$_{2.5}$
concentrations during the entire pregnancy.

Primary meta-analysis. According to the
common protocol, ICAPPO centers initially
estimated the association between term LBW
and maternal exposure to PM$_{10}$ averaged over
the entire pregnancy. Each center constructed
three logistic regression models to estimate the
odds of term LBW associated with each
10-µg/m$^3$ increase in average PM$_{10}$ exposure
levels during the entire pregnancy: a) without
any adjustment, b) with adjustment only for
maternal SES, and c) with adjustment for
maternal SES and center-specific covariates
(e.g., maternal age, maternal ethnicity, mater-
nal smoking, parity, and infant sex). For these
center-specific covariates, there was no recom-
mendation in the ICAPPO protocol and the
centers had the flexibility to independently
choose the suitable ones according to their
settings. This selection of extra covariates used
by each center has been reported elsewhere
(Parker et al. 2011). The construction of these
three models was to evaluate the effect of
adjustment for the maternal SES and other
covariates on the combined effect estimates.

ICAPPO participants chose maternal edu-
cation as a common indicator of maternal SES
(Parker et al. 2011). If maternal education
data were not available, area-level measures of
SES were used (for the PAMPER (Particulate
Matter and Perinatal Events Research) study:
Townsend Deprivation Score; for the Sydney
study: Index of Relative Socioeconomic
Disadvantage; and for the Vancouver study:
percentage of women with postsecondary edu-
cation). The PAMPER study (Newcastle upon
Tyne, UK) provided exposure data only for
black smoke, which approximates particulate
matter with an aerodynamic diameter ≤ 4 µm
(PM$_{4}$) and has been shown to be a reasonable
surrogate for PM$_{10}$ (Muir and Laxen 1995).
The estimates of effect for the PAMPER study
were therefore analyzed alongside the studies
with PM$_{10}$ measures.

We used meta-analysis to estimate com-
bined ORs across the centers. Between-
center heterogeneity was quantified using the
between-center variance of effect estimates, $t^2$.
The statistical significance of between-center heterogeneity was tested by Cochran’s Q test.
The $F$ statistic with 95% confidence intervals
(CI) was used to estimate the proportion of total
variation in effect estimates across centers
that was attributable to the between-center hetero-
genesis ($t^2$) rather than within-center error
(Higgins 2008). If there was significant
between-center heterogeneity (i.e., Cochran’s
Q test $p$-value < 0.05), DerSimonian–Laird
random-effects models were used for meta-
analysis (DerSimonian and Laird 1986); oth-
erwise, fixed-effects models were conducted
using the Mantel–Haenszel method (Mantel
and Haenszel 1959). Associations with a
Cochran’s Q test $p$-value between 0.05 and
0.10 were estimated using both fixed- and
random-effects models (Higgins et al. 2002).

Additional meta-analyses. Exposure win-
dow period. Nine centers analyzed the asso-
ciation between maternal exposure to PM$_{10}$
and term LBW stratified by the trimester of
exposure and adjusted for maternal SES. We
conducted meta-analyses using these reported
stratified ORs to evaluate the impact of the
exposure window period on the association between PM$_{10}$ exposure and term LBW.

**Birth weight as a continuous variable.** A subset of 11 centers (all centers but Connecticut and Massachusetts, New Jersey, and Seattle, Washington) estimated the change in term birth weight as a continuous outcome variable (grams) associated with each 10-µg/m$^3$ increase in PM$_{10}$ exposure levels averaged over the entire pregnancy using three sets of linear regression models with predictors as described for the primary analysis. The combined effect estimates across the centers were calculated using meta-analysis, as described above.

**PM$_{2.5}$ exposure.** Seven centers had data on PM$_{2.5}$ (Table 1) and reported the odds of term LBW associated with each 10-µg/m$^3$ increase in average PM$_{2.5}$ exposure levels during the entire pregnancy, using three sets of logistic regression models as described for the primary analysis. We synthesized the effect estimates across these centers using meta-analysis as described below.

**Sensitivity analyses.** We checked the robustness of all meta-analyses results to the omission of influential centers that were identified using DFBETAS, which indicates the change in estimated coefficients after excluding a center from the meta-analysis (Viechtbauer and Cheung 2010). Centers with an absolute DFBETAS value > 1 (i.e., that resulted in ≥ 1-SD change in the estimated coefficient when omitted) were considered influential (Viechtbauer and Cheung 2010).

**Meta-regressions.** To assess between-center heterogeneity, we hypothesized that the following characteristics of the centers could have an impact on the estimated risk by each center: study area (square kilometers), length of the study (years), number of births, continent of the study location, latitude of the study location (degrees), percentage of term LBW births, median PM$_{10}$ exposure levels (micrograms per cubic meter), median PM$_{2.5}$ exposure levels (micrograms per cubic meter), interquartile range of PM$_{10}$ exposure levels (micrograms per cubic meter), and PM$_{2.5}$/PM$_{10}$ ratio. In addition, we hypothesized that the use of model-based versus monitor-based exposure assessment, and temporal versus spatiotemporal exposure contrasts could influence center-specific estimates. Temporal exposure contrasts account for temporal variation in pollutant levels by assigning measurements from a single monitoring station to all study subjects, whereas spatiotemporal contrasts account for both spatial and temporal components of variation in air pollution levels when estimating exposure.

For associations between maternal PM$_{10}$ exposure during the entire pregnancy and term LBW that showed between-center heterogeneity, we performed separate univariate meta-regressions using center-specific log-transformed ORs as outcome and each of the center or exposure assessment characteristics listed above as predictors. In effect, these meta-regressions quantified the potential impact of these factors on the estimates of effects (ORs) reported by each center. We excluded the Generation R study (Table 1) from the analysis comparing temporal and spatiotemporal exposure contrasts because its exposure assessment was based on a dispersion model, which is essentially a spatial approach without any temporal adjustment. Analyses of the influence of PM$_{2.5}$ and PM$_{2.5}$/PM$_{10}$ ratios were based on the seven centers with PM$_{2.5}$ data.

**Results**

**Characteristics of ICAPPO centers.** Of the 14 ICAPPO centers included in our analyses (Table 1), 6 were North American, 5 European, 1 South American, 1 Asian,

### Table 1. Exposure assessment methodologies and characteristics of the ICAPPO centers.

<table>
<thead>
<tr>
<th>Center, location</th>
<th>Reference</th>
<th>Study period</th>
<th>Study area (km$^2$)</th>
<th>No. of births</th>
<th>Measure of SES</th>
<th>Term LBW (%)</th>
<th>Median PM$_{10}$ (µg/m$^3$)</th>
<th>Median PM$_{2.5}$ (µg/m$^3$)</th>
<th>Median PM$<em>{2.5}$/PM$</em>{10}$ ratio</th>
<th>Exposure assessment</th>
<th>Exposure contrast</th>
</tr>
</thead>
<tbody>
<tr>
<td>Atlanta, GA, USA</td>
<td>Darrow et al. 2011</td>
<td>1996–2004</td>
<td>4,538</td>
<td>325,221</td>
<td>Maternal education</td>
<td>2.62</td>
<td>23.5</td>
<td>3.1</td>
<td>15.8</td>
<td>0.67</td>
<td>Monitor</td>
</tr>
<tr>
<td>California, USA</td>
<td>Morello-Frosch et al. 2010</td>
<td>1996–2006</td>
<td>423,970</td>
<td>1,714,509</td>
<td>Maternal education</td>
<td>2.43</td>
<td>28.9</td>
<td>16.1</td>
<td>16.5</td>
<td>0.57</td>
<td>Monitor</td>
</tr>
<tr>
<td>Connecticut and Massachusetts, USA</td>
<td>Bell et al. 2007, 2008</td>
<td>1999–2002</td>
<td>41,892</td>
<td>173,042</td>
<td>Maternal education</td>
<td>2.16</td>
<td>22.0</td>
<td>7.4</td>
<td>20.0</td>
<td>0.91</td>
<td>Monitor</td>
</tr>
<tr>
<td>EDEN, Poitiers and Nancy, France</td>
<td>Lepeule et al. 2010</td>
<td>2003–2006</td>
<td>480</td>
<td>1233</td>
<td>Age at completion of education</td>
<td>2.11</td>
<td>19.0</td>
<td>3.0</td>
<td>—</td>
<td>—</td>
<td>Monitor</td>
</tr>
<tr>
<td>Lombardy, Italy</td>
<td>Pesatori et al. 2008</td>
<td>2004–2006</td>
<td>23,865</td>
<td>213,542</td>
<td>Maternal education</td>
<td>2.71</td>
<td>49.0</td>
<td>10.0</td>
<td>—</td>
<td>—</td>
<td>Monitor</td>
</tr>
<tr>
<td>PAMPER, Newcastle upon Tyne, UK</td>
<td>Gliniana et al. 2008</td>
<td>1962–1992</td>
<td>63</td>
<td>81,953</td>
<td>Area-level indicator$^b$</td>
<td>3.19</td>
<td>32.9$^c$</td>
<td>87.5$^c$</td>
<td>—</td>
<td>—</td>
<td>Model</td>
</tr>
<tr>
<td>New Jersey, USA</td>
<td>Rich et al. 2009</td>
<td>1999–2003</td>
<td>22,592</td>
<td>87,281</td>
<td>Maternal education</td>
<td>2.75</td>
<td>28.0</td>
<td>6.9</td>
<td>13.7</td>
<td>0.49</td>
<td>Monitor</td>
</tr>
<tr>
<td>PIAMA, North, West, and Center of the Netherlands</td>
<td>Gehring et al. 2011</td>
<td>1996–1997</td>
<td>12,000</td>
<td>3,471</td>
<td>Maternal education</td>
<td>1.15</td>
<td>40.5</td>
<td>6.7</td>
<td>20.3</td>
<td>0.50</td>
<td>Model</td>
</tr>
<tr>
<td>São Paulo, Brazil</td>
<td>Gouveia et al. 2004</td>
<td>2005</td>
<td>1,500</td>
<td>158,791</td>
<td>Maternal education</td>
<td>3.77</td>
<td>40.3</td>
<td>2.9</td>
<td>—</td>
<td>—</td>
<td>Monitor</td>
</tr>
<tr>
<td>Vancouver, BC, Canada</td>
<td>Brauer et al. 2008</td>
<td>1999–2002</td>
<td>3,300</td>
<td>66,467</td>
<td>Area-level indicator$^b$</td>
<td>1.35</td>
<td>12.5</td>
<td>1.4</td>
<td>3.98</td>
<td>0.32</td>
<td>Monitor</td>
</tr>
</tbody>
</table>

$^a$Ratio of PM$_{2.5}$ and PM$_{10}$ median levels. $^b$The Townsend Deprivation Score is an area-based measure of material deprivation calculated for each enumeration district (~ 200 households) based on 1971, 1981, and 1991 census data. $^c$Black smoke (~ PM$_{10}$) was used as a measure of particulate air pollution. $^d$The Australian Bureau of Statistics (2003) Index of Relative Socio-economic Disadvantage uses a range of census factors and is assigned to each census collection district (~ 200 households). $^e$The percentage of women with postsecondary education.

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and 1 was Oceanian. Our analysis included > 3 million births (ranging from a little more than 1,000 to almost 2 million) generally occurring between late 1990s and mid-2000s.

Primary meta-analysis. Thirteen centers provided estimates for the association between maternal exposure to PM$_{10}$ and term LBW [Figure 1A; see also Supplemental Material, Figure S1, for forest plots (http://dx.doi.org/10.1289/ehp.1205575)]. There was statistically significant (Cochran’s Q test p-value < 0.05) between-center heterogeneity in effect estimates (ORs) reported by these centers, with $\tau^2$ ranging between 0.0003 and 0.0004 (Table 2). Therefore we used random-effects models to estimate combined ORs across the centers, which indicated a positive association between term LBW and average maternal exposure to PM$_{10}$ during the entire pregnancy before adjustment (OR = 1.04; 95% CI: 1.01, 1.06) and after adjustment for SES (OR = 1.03; 95% CI: 1.01, 1.05) and SES plus center-specific covariates (OR = 1.02; 95% CI: 1.01, 1.04) (Table 2).

Birth weight as a continuous variable. Eleven centers estimated the change in term birth weight as a continuous outcome variable (grams) associated with each 10-µg/m$^3$ increase in PM$_{10}$ exposure levels [Figure 1B; see also Supplemental Material, Figure S3, for forest plots (http://dx.doi.org/10.1289/ehp.1205575)]. The Cochran’s Q test of heterogeneity was significant ($p < 0.05$) for all meta-analyses of PM$_{10}$ and birth weight (data not shown). Random-effects meta-analyses indicated a negative association between term birth weight and a 10-µg/m$^3$ increase in PM$_{10}$, with unadjusted, SES-adjusted, and SES- plus center-specific covariate–adjusted estimated decreases of –2.7 g (95% CI: –7.2, 1.7 g, $p = 0.23$), –3.0 g (95% CI: –6.9, 0.9 g, $p = 0.13$), and –8.9 g (95% CI: –13.2, –4.6 g, $p < 0.01$), respectively.

PM$_{2.5}$ exposure. Seven centers reported the odds of term LBW associated with each 10-µg/m$^3$ increase in PM$_{2.5}$ exposure levels [Figure 1C; see also Supplemental Material, Figure S4, for forest plots (http://dx.doi.org/10.1289/ehp.1205575)]. All meta-analyses showed statistically significant between-study heterogeneity (Table 2). The random-effects meta-analyses demonstrated positive associations with term LBW, with statistically significant ORs based on unadjusted and SES-adjusted models, but not the model adjusted for maternal SES and center-specific covariates (Table 2).

Sensitivity analyses. The studies that were most frequently classified as influential were the California study (the largest center with more than 1.7 million births), the PAMPER study (covering a long time period with wide variation in sources and levels of exposure to black smoke, which was used as a surrogate for PM$_{10}$), and the Lombardy study (a relatively large study of a heterogeneous region including a metropolitan area...
(Milan), a large agricultural area (Po valley), and a northern mountainous area [see Supplemental Material, Table S1 (http://dx.doi.org/10.1289/ehp.1205575)]. The meta-analyses were generally robust to the exclusion of influential studies with regard to the magnitude of the estimated associations (data not shown), but there were some exceptions. For example, the unadjusted OR for LBW with a 10-µg/m³ increase in PM_{2.5} increased from 1.17 (95% CI: 1.08, 1.26) to 1.20 (95% CI: 1.00, 1.45) after California was excluded, and decreased to 1.06 (95% CI: 0.99, 1.13) after exclusion of Connecticut and Massachusetts. In addition, the average estimated reduction in SES-adjusted term birth weight with a10-µg/m³ increase in PM_{10} (random-effects model) increased from −3.0 g (95% CI: −6.9, 0.9 g) based on all centers (Figure 1B) to −5.5 g (95% CI: −9.3, −1.6 g) after excluding Atlanta, and to −4.9 g (95% CI: −8.6, −1.1 g) after excluding Lombardy. Similarly, the reduction in unadjusted mean birth weight increased from −2.7 g (95% CI: −7.2, 1.7 g) based on all centers (see Supplemental Material, Figure S3A) to −6.6 g (95% CI: −11.7, −1.5 g) and −5.0 g (95% CI: −10.0, 0.0 g) after removing Atlanta and Lombardy, respectively.

Meta-regressions. Of the characteristics we evaluated, median PM_{2.5} exposure levels, PM_{2.5}/PM_{10} ratio, and exposure contrast (temporal vs. spatiotemporal approach) influenced the between-center heterogeneity in the reported PM_{2.5}-term LBW associations, with centers that had higher median PM_{2.5} exposure levels and PM_{2.5}/PM_{10} ratios, and that used a temporal exposure contrast, reporting stronger associations in most cases (Table 4). The results of leave-one-out sensitivity analyses for these three meta-regressions were generally consistent with those of meta-regressions including all centers (data not shown). However, removing Connecticut and Massachusetts almost doubled the meta-regression coefficient for the PM_{2.5}/PM_{10} ratio and nullified the association for median PM_{2.5} exposure levels, and the meta-regression estimate for median PM_{2.5} exposure levels was null after New Jersey was excluded.

Meta-regressions for the exposure contrast showed statistically significant heterogeneity in residuals (from both unadjusted and SES-adjusted models), whereas meta-regressions of the median PM_{2.5} exposure levels and PM_{2.5}/PM_{10} ratios did not (Table 4). Heterogeneity was reduced when ORs were adjusted for center-specific selection of covariates in addition to SES. Associations between center-specific log-ORs and other center or study characteristics were not statistically significant (data not shown).

Discussion

To our knowledge, this study is the largest multicenter study so far reporting on the association between air pollution and fetal growth using a common analytical protocol. We synthesized reported estimates of effects by 14 ICAPPO centers around the globe and for the first time quantified the impacts of study settings and aspects of applied exposure assessment methods on between-center heterogeneity in the reported effect estimates (ORs) for this association. We found that maternal exposures to PM_{10} and PM_{2.5} during the entire pregnancy were positively associated with term LBW. For PM_{10}, all trimester-specific exposures were associated with slightly increased odds of term LBW. Furthermore, term birth weight was reduced in association with average PM_{10} exposure over the entire pregnancy. Most associations showed between-center heterogeneity in the center-specific estimates of associations. Meta-regression assessments of factors possibly affecting PM_{10}-term LBW associations indicated that heterogeneity was influenced by median PM_{2.5} exposure levels, the PM_{2.5}/PM_{10} ratio, and the applied exposure contrast (temporal vs. spatiotemporal).

Primary meta-analyses. Making decisions about the risks from environmental exposure in the policy or clinical setting requires

Table 2. Combined random-effects ORs (95% CIs) for term LBW in association with a 10-µg/m³ increase in average maternal exposure to PM_{2.5} and PM_{10} during pregnancy, and corresponding indicators of between-center heterogeneity across ICAPPO centers.

<table>
<thead>
<tr>
<th>Meta-analysis</th>
<th>Combined estimate</th>
<th>Heterogeneity</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>OR (95% CI)</td>
<td>p-Value</td>
</tr>
<tr>
<td>PM_{2.5} (13 centers)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Unadjusted</td>
<td>1.04 (1.01, 1.06)</td>
<td>&lt;0.01</td>
</tr>
<tr>
<td>Adjusted for maternal SES</td>
<td>1.03 (1.01, 1.05)</td>
<td>&lt;0.01</td>
</tr>
<tr>
<td>Adjusted for maternal SES and center-specific covariates</td>
<td>1.02 (1.01, 1.04)</td>
<td>0.01</td>
</tr>
<tr>
<td>PM_{2.5} (7 centers)b</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Unadjusted</td>
<td>1.17 (1.08, 1.26)</td>
<td>&lt;0.01</td>
</tr>
<tr>
<td>Adjusted for maternal SES</td>
<td>1.10 (1.03, 1.18)</td>
<td>&lt;0.01</td>
</tr>
<tr>
<td>Adjusted for maternal SES and center-specific covariates</td>
<td>1.04 (0.99, 1.09)</td>
<td>0.09</td>
</tr>
</tbody>
</table>

*p-Value for Cochran’s Q test for heterogeneity. bIncluding Atlanta, California, Connecticut and Massachusetts, New Jersey, PIAMA, Seattle, and Vancouver.

Table 3. Combined adjusteda ORs (95% CIs) of term LBW in association with a 10-µg/m³ increase in average maternal exposure to PM_{10} during pregnancy, and corresponding indicators of between-center heterogeneity across ICAPPO centers.

<table>
<thead>
<tr>
<th>Trimester of exposure</th>
<th>Combined risk</th>
<th>Heterogeneity</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>OR (95% CI)</td>
<td>p-Value</td>
</tr>
<tr>
<td>1st trimester (fixed-effects model)</td>
<td>1.01 (1.01, 1.01)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>1st trimester (random-effects model)</td>
<td>1.00 (1.00, 1.01)</td>
<td>0.325</td>
</tr>
<tr>
<td>2nd trimester (fixed-effects model)</td>
<td>1.01 (1.01, 1.02)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>2nd trimester (random-effects model)</td>
<td>1.01 (1.01, 1.02)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>3rd trimester (fixed-effects model)</td>
<td>1.02 (1.01, 1.02)</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>3rd trimester (random-effects model)</td>
<td>1.01 (1.01, 1.02)</td>
<td>0.001</td>
</tr>
</tbody>
</table>

*aAdjusted for maternal SES. bORs from nine centers were included in the meta-analysis. c*p-Value for Cochran’s Q test for heterogeneity.

Table 4. Meta-regression coefficients (95% CIs) from separate models estimating the effect of a 1-µg/m³ increase in the center median PM_{2.5} level, a 100% increase in PM_{2.5}/PM_{10} ratio, or the use of a temporal versus spatiotemporal exposure contrast on SES-adjusted center-specific log-ORs for the association between a 10-µg/m³ increase in mean PM_{2.5} during pregnancy and term LBW.

<table>
<thead>
<tr>
<th>Meta-regression</th>
<th>Temporal vs. spatiotemporal approacha,b</th>
<th>Median PM_{2.5} levels</th>
<th>PM_{2.5}/PM_{10} ratio</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Regression coefficient (95% CI)</td>
<td>Residual heterogeneity (95% CI)</td>
<td>Regression coefficient (95% CI)</td>
</tr>
<tr>
<td>Unadjusted</td>
<td>0.07 (0.03, 0.11)</td>
<td>0.0033 (0.0007, 0.0259)</td>
<td>0.03 (0.01, 0.05)</td>
</tr>
<tr>
<td>Adjusted for maternal SES</td>
<td>0.08 (0.05, 0.12)</td>
<td>0.0015 (0.0001, 0.0116)</td>
<td>0.02 (0.00, 0.04)</td>
</tr>
<tr>
<td>Adjusted for maternal SES and center-specific covariates</td>
<td>0.06 (0.02, 0.10)</td>
<td>0.0006 (0.0, 0.0079)</td>
<td>0.01 (0.00, 0.03)</td>
</tr>
</tbody>
</table>

aRegression coefficients for using temporal approach compared with the spatiotemporal approach. bGeneration R cohort was excluded because its exposure assessment was based on a dispersion model which is a spatial approach. cCochran’s Q test p < 0.05.
synthesizing and interpreting the available epidemiologic evidence, which may include evaluating a number of relevant studies with different designs. A recent systematic review and meta-analysis of the available evidence on the association between particulate pollution and pregnancy outcomes reported by Sapkota et al. (2010) estimated a combined OR of 1.02 (95% CI: 0.99, 1.05) for LBW in association with a 10-µg/m³ increase in average maternal PM₁₀ exposure during pregnancy. Our estimates, which were based on center-specific ORs estimated using a common protocol, were slightly more precise than reported by Sapkota et al. (e.g., OR = 1.03; 95% CI: 1.01, 1.05 for a 10-µg/m³ increase in average maternal PM₁₀ exposure based on SES-adjusted estimates); but both point estimates support a comparable deleterious effect of PM₁₀ on term birth weight, an indicator of fetal growth. This comparable finding supports not overrelying on statistical significance nor using it as deciding factor in the evaluation of the evidence, but rather describing the degree of confidence and precision of the estimate, which is consistent with other scientific writings on this topic. Bacchetti et al. (2005, 2008), for example, caution against inappropriate reliance on statistical significance and rather support the decision-making process evaluating the degree of confidence in the findings and the effect size. As we have shown by comparing our results to those of Sapkota et al. (2010), even though we were able to improve the signal-to-noise ratio using our approach, the interpretation of the relationship between air pollution and fetal growth remained similar.

The consistency of our findings with those of Sapkota et al. (2010), our observed consistency of combined ORs and between-center heterogeneity (τ² values) for meta-analyses of ORs adjusted for center-specific covariates, and our meta-analyses of unadjusted or SES-adjusted ORs give us confidence that synthesizing effect estimates for the air pollution–pregnancy outcomes associations reported by studies with different designs is informative. Further, experience from this study could be important for policy makers in incorporating research evidence into policy—for example, by including estimates of fetal growth in the future reviews of air quality standards.

ICAPPO centers conducted unadjusted and SES-adjusted analyses according to a common analysis protocol, as well as analyses that allowed center-specific selection of covariates. In combining results across centers, there were clear trade-offs to the two approaches. Strictly specifying the models and covariates to be consistent across centers was one way to reduce methodological differences between centers. However, there might be important center-specific considerations that required inclusion of certain covariates for validity. For example, the Atlanta results relied entirely on temporal contrasts of exposure over an 11-year time period, and failing to control for long-term time trends, as was done in the unadjusted and SES-adjusted analyses, may have led to temporal confounding. This might explain why ORs adjusted for center-specific selection of covariates were different from those unadjusted or adjusted for SES only in some cases.

Additional meta-analyses. In their meta-analysis, Sapkota et al. (2010) estimated the association between LBW and maternal PM₁₀ exposure during the first trimester (five studies) and third trimester (seven studies), but did not detect associations with exposure in either trimester (OR = 1.00; 95% CI: 0.97, 1.03, and OR = 1.00; 95% CI: 0.99, 1.01, respectively). We estimated slightly increased relative risks of term LBW in association with PM₁₀ exposure during all trimesters, consistent with other previous reports (Parker and Woodruff 2008; Parker et al. 2005).

Because Cochran’s Q test of heterogeneity p-values for the trimester-specific analyses of PM₁₀ and LBW were between 0.05 and 0.10, we reported results of both fixed-effects and random-effects meta-analyses (Higgins et al. 2002). For the second and third trimesters, there were no notable differences in estimates between the two models, whereas the random-effects OR for the association with first-trimester PM₁₀ was null. Given the evidence of heterogeneity, fixed-effects model estimates for this trimester should be interpreted with caution.

Combined ORs for term birth weight as a continuous variable indicated a reduction in term birth weight associated with a 10-µg/m³ increase in average PM₁₀ exposure during the entire pregnancy (–2.7 g; 95% CI: –7.2, 1.7 g). The association was stronger when based on ORs adjusted for maternal SES and center-specific covariates (–8.9 g; 95% CI: –13.2, –4.6 g). This may be partly explained by adjustment for gestational age at delivery by nine centers. Gestational age at delivery could confound the association between maternal exposure to air pollution and fetal growth. Although the outcome was term birth weight or LBW, there could be a 6-week difference (between 37 and 42 weeks of gestation) in gestational age at delivery. Confounding by gestational age might have had a stronger effect on the continuous birth weight analyses than on the LBW analyses.

LBW was associated with maternal PM₂·₅ exposure (combined random-effects OR = 1.10; 95% CI: 1.03, 1.18 for a 10-µg/m³ increase in average PM₂·₅ based on SES-adjusted ORs). The strength and direction of this association was comparable with the meta-analysis OR reported by Sapkota et al. (2010) (OR = 1.09; 95% CI: 0.90, 1.32).

Meta-regressions. To our best knowledge, the impact of study characteristics and exposure assessment methods on estimated associations between adverse birth weight and maternal exposure to air pollution has not been evaluated previously. Median PM₂·₅ exposure levels, PM₂·₅:PM₁₀ ratios, and temporal versus spatiotemporal exposure contrasts appeared to influence the estimates of effects reported by the study centers. Parker and Woodruff (2008) suggested that variation in the composition of particulate matter may have contributed to differences in associations between maternal exposure to particulate pollution and fetal growth among seven different regions in the United States (Parker and Woodruff 2008). However, we did not have data on the composition of particulate pollutants, and therefore were not able to investigate its impact.

In our analysis, meta-regressions of the influence of temporal versus spatiotemporal exposure contrasts on associations between PM₁₀ and LBW generally showed strong evidence of residual heterogeneity, in contrast to meta-regressions of median PM₂·₅ exposure levels and PM₂·₅:PM₁₀ ratios (Table 4). These findings could suggest that at least part of the effect of the exposure contrast on PM₁₀-term LBW associations might have been secondary to the effect of PM₂·₅:PM₁₀ ratios and median PM₂·₅ levels at each center.

Limitations. Our meta-regressions were based on effect estimates (ORs) from 13 ICAPPO centers. As a rule of thumb, it has been suggested that 10 effect estimates are required in order to include a covariate in meta-regression (Borenstein et al. 2009). We therefore had to limit our analysis to univariate rather than multivariate meta-regressions and test the effects of each covariate separately. As a result, we could not evaluate whether our univariate results were confounded by other factors. Furthermore, other center characteristics that could have affected center-specific effect estimates, such as the prevalence of maternal smoking, were not included in our meta-regression analyses.

The ICAPPO protocol did not include data that might have been used to better characterize personal exposure, such as direct measurements of personal exposure levels (e.g., personal monitor data), pollutant levels at microenvironmental levels (e.g., indoor, outdoor, commuting) or maternal time-activity patterns. Some exposure misclassification would have resulted from the use of effect estimates based on associations with ambient levels of pollutants as a surrogate for personal exposure levels.

Conclusion. Our combined effect estimates, which were based on effect estimates generated by 14 ICAPPO centers across the globe using a common analytical protocol, support an adverse impact of maternal exposure to particulate pollution on fetal growth. The estimated
combined associations, although relatively small, could be of major public health importance considering the ubiquitous nature of particulate air pollution exposure and therefore the potential for considerable population attributable risks, particularly given evidence of both perinatal and lifelong effects of LBW on health (Balci et al. 2010; Gibson 2007). After reducing analytical differences as a possible source of heterogeneity by using a common protocol, we found that some of the heterogeneity in effect estimates by centers could be explained by differences in median \( PM_{2.5} \) exposure levels and \( PM_{2.5}/PM_{10} \) ratios, suggesting geographical variation in the association between air pollution and fetal growth. In general, the direction and strength of combined estimates of association and between-center heterogeneity based on unadjusted and SES-adjusted ORs were consistent with combined estimates based on ORs adjusted for center-specific covariates. These findings highlight the contribution of study settings to inconsistencies in the available literature and can therefore increase the confidence of policy makers when summarizing existing evidence and translating it into policy.

**References**


