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#### Full length article

# A nationwide study of maternal exposure to ambient ozone and term birth weight in the United States

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#### ABSTRACT

Background: Maternal exposure to ozone  $(O_3)$  may cause systemic inflammation and oxidative stress and contribute to fetal growth restriction. We sought to estimate the association between maternal exposure to  $O_3$  and term birth weight and term small for gestational age (SGA) in the United States (US).

Methods: We conducted a nationwide study including 2,179,040 live term singleton births that occurred across 453 populous counties in the contiguous US in 2002. Daily county-level concentrations of  $O_3$  data were estimated using a Bayesian fusion model. We used linear regression to estimate the association between  $O_3$  exposure and term birth weight and logistic regression to estimate the association between  $O_3$  exposure and term SGA during each trimester of the pregnancy and the entire pregnancy after adjusting for maternal characteristics, infant sex, season of conception, ambient temperature, county poverty rate, and census region. We additionally used distributed lag models to identify the critical exposure windows by estimating the monthly and weekly associations.

Results: A 10 parts per billion (ppb) increase in  $O_3$  over the entire pregnancy was associated with a lower term birth weight (-7.6 g; 95 % CI: -8.8 g, -6.4 g) and increased risk of SGA (odds ratio = 1.030; 95 % CI: 1.020, 1.040). The identified critical exposure windows were the 13th- 25th and 32nd -37th gestational weeks for term birth weight and 13th- 25th for term SGA.

We found the association was more pronounced among mothers who were non-Hispanic Black, unmarried, or had lower education level.

Conclusions: Among US singleton term births, maternal exposure to  $O_3$  was associated with lower rates of fetal growth, and the 13th- 25th gestational weeks were the identified critical exposure windows.

#### 1. Introduction

Birth weight is a good proxy for infant morbidity and mortality and may also influence the health status in adulthood (Horikoshi et al. 2016; Zhang et al. 2014). Known risk factors for fetal growth restriction include genetic makeup, placental insufficiency, poor nutrition, and

maternal morbidities, such as chronic hypertension and gestational diabetes (Malhotra et al. 2019).

Previous studies suggested that environmental risk factors might adversely affect fetal growth, such as ambient heat (Sun et al. 2019a) and ambient air pollution (Bekkar et al. 2020; Li et al. 2017). For example, synthesizing evidence from 32 studies, a recent *meta*-analysis

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reported that each  $10~\mu\text{g/m}^3$  increase in full pregnancy fine particulate matter (PM<sub>2.5</sub>) exposure led to a reduction in birth weight by -15.9~g (95 % CI: -26.8~g, -5.0~g) (Sun et al. 2016). Ozone (O<sub>3</sub>) is one of the criteria air pollutants and is associated with a broad range of health outcomes, such as asthma (Dai et al. 2018) and cardiovascular diseases (Niu et al. 2022). Maternal exposure to O<sub>3</sub> could trigger preterm births (Rappazzo et al. 2021), but the evidence for birth weight is limited (Chen et al. 2002; Gouveia et al. 2004; Lamichhane et al. 2020; Morello-Frosch et al. 2010; Niu et al. 2020; Stieb et al. 2012; Wang et al. 2021). Fetal growth requires adequate oxygen and nutrient; exposure to O<sub>3</sub> may affect placental functions by inducing inflammation and oxidative stress, leading to reduced birth weight (EPA 2020).

Leveraging a nationwide study of  $\sim 2.2$  million US live term births, we aimed to estimate the association of maternal exposure to  $O_3$  with term birth weight and term small for gestational age (SGA). We also sought to identify susceptible exposure windows and to examine whether the associations varied by maternal characteristics.

#### 2. Methods

#### 2.1. Study population

We obtained data on US live births from the CDC's National Center for Health Statistics (NCHS 2022). For each birth, we obtained data on county of maternal residence, gestational weeks, maternal age, maternal race/ethnicity, maternal educational attainment, marital status, parity, smoking and alcohol drinking during pregnancy, chronic hypertension, date of last menstrual period (LMP), and infant sex. Given the availability of air pollution data, county of maternal residence, and datacoding consistency between years, we used the 2002 birth data only. We included only singleton births (i.e., a child that is the only one born at birth) with gestational age between 37 and 44 weeks and birth weight between 1000 g and 5500 g born to mothers who resided in counties with populations of 100,000 or more in the contiguous US. We excluded births with implausible combinations of birth weight and gestational age based on the Alexander criteria (Alexander et al. 1996). We excluded births with missing values of race/ethnicity, parity, educational attainment, beginning month of prenatal care, and LMP (Fig S1). We did not exclude births with missing values of smoking, alcohol drinking, or chronic hypertension, as a large proportion of these variables were recorded as unknown or not stated. We finally included 2,179,040 term births in our analytical data set.

#### 2.2. Birth outcome definition

The birth outcomes were term birth weight and term small for gestational age (SGA). We obtained birth weight in grams and gestational weeks based on LMP from the birth certificate. We chose term SGA as it takes gestational length into account. We used the sex- and gestational week-specific US national reference percentiles for birth weight to classify each birth into corresponding birth weight percentiles (Oken et al. 2003), and those births with birth weight less than the 10th percentile were classified as SGA.

#### 2.3. Air pollution assessment

Daily concentrations of 8-hour maximum  $O_3$  and 24-hour average  $PM_{2.5}$  with spatial resolutions of 36 km  $\times$  36 km and 12 km  $\times$  12 km were obtained from the US Environmental Protection Agency (EPA) for the Tracking Network. These data were generated using hierarchical Bayesian space–time downscaling fusion models combining output from the Community Multi-scale Air Quality model with air quality monitoring data from US EPA's air quality system (Berrocal et al. 2012; Byun and Schere 2006). The daily county-level population-weighted  $O_3$  and  $PM_{2.5}$  data were then aggregated by summing the predictions of air pollution concentrations falling within that census tract based on the

2000 US Census multiplying the corresponding proportion of the county population. Details about the theory and evaluation of the downscaler model have been published elsewhere (Berrocal et al. 2010; Berrocal et al. 2012). These data have been used to estimate the health effects of air pollution in the US (Mirabelli et al. 2016; Strosnider et al. 2019). Based on the LMP, we calculated weekly, monthly, and trimester average O<sub>3</sub> exposure for each birth and also calculated average PM<sub>2.5</sub> exposure during the full pregnancy for each pregnancy (eMethods).

#### 2.4. Ambient temperature assessment

We obtained gridded daily ambient temperature estimates from the Parameter-elevation Relationships on Independent Slopes model and calculated daily population-weighted average of ambient temperature for each county (Spangler et al. 2019; Sun et al. 2019a; Sun et al. 2019b). We then calculated the average ambient temperature during the full pregnancy for each birth (eMethods).

#### 2.5. Statistical analysis

We used linear regression to estimate the association between O<sub>3</sub> exposure and term birth weight and logistic regression to estimate the association between O<sub>3</sub> exposure and term SGA during each trimester and the full pregnancy. In the main models, we adjusted for maternal age (18-24, 25-29, 30-34, 35-40, 40-50 years), maternal race/ ethnicity (non-Hispanic White, non-Hispanic Black, other), maternal educational attainment (below high school, high school, above high school), marital status (married versus not married), parity (0, 1, >2), infant sex (male versus female), chronic hypertension (yes, no, unknown), alcohol drinking during pregnancy (yes, no, unknown), smoking during pregnancy (yes, no, unknown), prenatal care initiation time (no prenatal care, 1st-3rd month, 4th-6th month, 7th month to final month), conception season (spring: March-May, summer: June-August, fall: September-November, winter: December-February]), county poverty rate defined as the percentage of 2002 county residents below the federal poverty line (<10%, 10%-14.9%, 15%-19.9%,  $\ge 20\%$ ) (U. S. Census Bureau 2022), census regions (Northeast, Midwest, South, West), mean ambient temperature during the pregnancy using natural cubic splines with 3 degrees of freedom, and mutually adjusted for O<sub>3</sub> exposure in other trimesters. We adjusted maternal race/ethnicity and census regions in the main models as the level of air pollution exposure might vary by race/ethnicity and geographic regions (Jbaily et al. 2022). For term birth weight models, we additionally adjusted for gestational age (37-38, 39-40, 41-42, 43-44 weeks).

We conducted sensitivity analyses by additionally adjusting for PM $_{2.5}$  to assess potential confounding by other air pollutants. As the causal relationships between ambient temperature,  $O_3$ , and disease outcomes are complex (Buckley et al. 2014), we conducted sensitivity analyses without adjustment for ambient temperature during the entire pregnancy. We reported changes in birth weight (g) for term birth weight and odds ratios (ORs) for term SGA per 10 parts per billion (ppb) increase in  $O_3$  concentration.

To identify potential susceptible subgroups, we conducted subgroup analysis to assess whether the associations between full pregnancy  $O_3$  exposure and term birth weight or term SGA varied by maternal age, maternal race/ethnicity, maternal educational attainment, marital status, and infant sex. We performed a Wald statistic to assess the heterogeneity across subgroups (Rothman et al. 2008).

To identify the critical exposure window, we used distributed lag models combined with linear regression or logistic regression to estimate the monthly and weekly associations for term birth weight or term SGA. We modeled the association between O<sub>3</sub> exposure and term birth weight or term SGA using a linear function and modeled the lag function using a natural cubic function with knots placed at equal intervals on the scale of lags up to 37 gestational weeks (Wilson et al. 2017). We set the maximum lag as 37 gestational weeks to ensure an equal length of

exposure among all pregnant women. The number of knots for the lag function was selected to minimize the Akaike Information Criterion (Table S1).

We carried out all statistical analyses with R software (version: 3.5.1) and fitted the distributed lag models using the "dlnm" package (version: 2.4.7).

#### 3. Results

Our analysis included 2,179,040 singleton term births across 453 US counties in 2002, with an average birth weight of 3,423 g (Table 1 & Fig. 1). Among the population, 9.5 % of which were born SGA. The birth weight was lower among first live births or births born to mothers younger than 25 years, unmarried individuals, non-Hispanic Black individuals, individuals with a high school education, individuals who received no prenatal care, or smoked or consumed alcohol during pregnancy, or living in counties with higher level of poverty rate.

#### 3.1. Trimester O<sub>3</sub> exposure and fetal growth

Maternal exposure to O<sub>3</sub> was associated with a lower term birth weight and a higher risk of term SGA (Table 2). In the main models after adjusting for individual-level covariates, county-level poverty rate and census region, mean ambient temperature during pregnancy, and mutually adjusted for O<sub>3</sub> exposure in the other trimesters, a 10 ppb increase in O3 during the pregnancy was associated with a reduction in birth weight of -7.6 g (95 % CI: -8.8 g, -6.4 g) for term birth weight or an OR of 1.030 (95 % CI: 1.020, 1.040) for term SGA. The associations were also observed in all three trimesters for term birth weight and in the second and third trimesters for term SGA with the strongest association consistently observed in the second trimester. For example, the OR per 10 ppb increase in O<sub>3</sub> exposure for term SGA was 1.016 (95 % CI: 1.008, 1.024) during the second trimester, 1.006 (95 % CI: 1.000, 1.013) during the third trimester, and 1.005 (95 % CI: 0.998, 1.012) during the first trimester. Results were similar in sensitivity analyses with additionally adjusted for PM2.5 or without adjustment for the average ambient temperature during the entire pregnancy (Table 2).

#### 3.2. Subgroup analysis

We conducted subgroup analyses to assess whether the association of term birth weight or term SGA with full pregnancy O3 exposure varied by maternal age, maternal race/ethnicity, educational attainment, marital status, and infant sex. We found a statistically significant heterogeneity across race/ethnicity, educational attainment, and marital status (p for heterogeneity < 0.05) for both term birth weight and term SGA, with the strongest impact observed among non-Hispanic Black mothers, mothers with high school education, or unmarried mothers (Table 3). For example, the reduced birth weight per 10 ppb increase in  $O_3$  during pregnancy was -13.8 g (95 % CI: -15.1 g, -12.0 g) among non-Hispanic Black mothers, -6.7 g (95 % CI: -10.8 g, -3.0 g) among non-Hispanic White mothers, and -9.5 g (95 % CI: -11.8 g, -7.0 g) among Hispanic mothers, and the corresponding estimate for term birth weight was -13.8 g (95 % CI: -15.6 g, -12.0 g), -6.7 g (95 % CI: -10.8g, -3.0 g), and -9.5 g (95 % CI: -11.8 g, -1.0 g), respectively. We found the association was stronger among younger women for term birth weight but was not observed for term SGA. We found no evidence of any heterogeneity between male and female infants (p for heterogeneity > 0.05).

#### 3.3. Critical exposure windows

To identify potential critical exposure windows, we fitted distributed lag models to estimate associations for  $O_3$  exposure in each of the 9 months or 37 gestational weeks of pregnancy while including  $O_3$  concentrations in the other months or gestational weeks of pregnancy in the

**Table 1**Demographics of singleton term births in the United States in 2002.

Characteristics	Term births (%)	Term birth weight (g) mean (SD)	Term SGA (%) <sup>a</sup>	
Total population	2,179,040	3423 (471)	208,064	
	(100.0)		(9.5)	
Infant sex				
Male	1,108,434	3485 (476)	106,728	
	(50.9)		(9.6)	
Female	1,070,606	3359 (458)	101,336	
	(49.1)		(9.5)	
Parity				
0	863,624 (39.6)	3379 (467)	100,183	
			(11.6)	
1	728,342 (33.4)	3449 (463)	57,984 (8.0	
$\geq$ 2	587,074 (26.9)	3456 (482)	49,897 (8.5	
Gestational age (weeks)				
37–38	637,385 (29.3)	3266 (470)	57,549 (9.0	
39–40	1,138,418	3470 (450)	110,009	
	(52.2)		(9.7)	
41–42	340,158 (15.6)	3550 (468)	34,222	
			(10.1)	
43–44	63,079 (2.9)	3474 (475)	6284 (10.0)	
Maternal age (years)				
18–24	666,400 (30.6)	3350 (460)	83,145	
			(12.5)	
25–29	600,258 (27.5)	3431 (466)	54,070 (9.0	
30–34	576,647 (26.5)	3471 (470)	44,043 (7.6	
35–40	277,041 (12.7)	3477 (482)	21,558 (7.8	
40–50	58,694 (2.7)	3452 (499)	5248 (8.9)	
Maternal race/ethnicity				
Non-Hispanic White	1,166,173	3481 (467)	85,987 (7.4	
	(53.5)			
Non-Hispanic Black	302,817 (13.9)	3280 (472)	47,814	
			(15.8)	
Hispanic	553,674 (25.4)	3412 (461)	53,204 (9.6	
Other	156,376 (7.2)	3309 (447)	21,059	
			(13.5)	
Marital status				
Married	1,521,456	3458 (466)	120,584	
	(69.8)		(7.9)	
Not married	657,584 (30.2)	3341 (473)	87,480	
			(13.3)	
Maternal educational				
attainment (years)				
Lower than high school	132,851 (6.1)	3414 (469)	13,441	
			(10.1)	
High school	912,100 (41.9)	3378 (475)	106,091	
			(11.6)	
Higher than high school	1,134,089	3460 (465)	88,532 (7.8	
	(52.0)			
Timing of initiation of				
prenatal care				
No prenatal care	13,400 (0.6)	3264 (517)	2600 (19.4)	
1st-3rd month	1,866,327	3433 (469)	168,394	
	(85.6)		(9.0)	
4th-6th month	247,416 (11.4)	3372 (476)	30,076	
			(12.2)	
7th-final month	51,897 (2.4)	3353 (476)	6994 (13.5)	
Smoking during pregnancy				
Yes	153,211 (7.0)	3252 (473)	27,517	
			(18.0)	
No	1,652,027	3434 (468)	147,264	
	(75.8)		(8.9)	
Unknown <sup>b</sup>	373,802 (17.2)	3445 (468)	33,283 (8.9	
Alcohol drinking during				
pregnancy				
Yes	14,178 (0.7)	3355 (503)	2014 (14.2)	
No	1,788,486	3419 (471)	172,563	
	(82.1)	· ()	(9.6)	
Unknown <sup>b</sup>	376,376 (17.3)	3446 (469)	33,487 (8.9	
Chronic hypertension	, (27.10)	( )	,, (0.)	
Yes	14,517 (0.7)	3330 (546)	2140 (14.7)	
No	2,151,932	3423 (470)	204,749	
	(98.8)	0.20(1/0)	(9.5)	
Unknown	12,591 (0.6)	3452 (484)	1175 (9.3)	
CIMIOWII	12,071 (0.0)	J7J4 (704)	11/3 (9.3)	

(continued on next page)

Table 1 (continued)

Characteristics	Term births (%)	Term birth weight (g) mean (SD)	Term SGA (%) <sup>a</sup>
Season of conception <sup>c</sup>			
Spring	525,102 (24.1)	3420 (472)	51,321 (9.8)
Summer	537,690 (24.7)	3431 (472)	50,066 (9.3)
Fall	563,279 (25.8)	3423 (470)	53,721 (9.5)
Winter	552,969 (25.4)	3418 (469)	52,956 (9.6)
County poverty rated			
<10 %	799,652 (36.7)	3456 (470)	66,666 (8.3)
10 %-14.9 %	886,297 (40.7)	3414 (470)	87,148 (9.8)
15 %-19.9 %	362,023 (16.6)	3394 (470)	38,493
			(10.6)
≥20 %	131,068 (6.0)	3361 (470)	15,757
			(12.0)
Census region			
Northeast	463,736 (21.3)	3437 (475)	43,976 (9.5)
Midwest	484,518 (22.2)	3440 (476)	44,550 (9.2)
South	637,221 (29.2)	3391 (468)	65,701
			(10.3)
West	593,565 (27.2)	3433 (466)	53,837 (9.1)

<sup>&</sup>lt;sup>a</sup> The percentage of term small for gestational age was obtained by dividing the number of term small for gestational age by the number of term births in the corresponding category.

models (Fig. 2). For term birth weight,  $O_3$  exposures during the 4th-6th, 8th, and 9th gestational months or the 13th-24th and 32nd-37th gestational weeks were associated with reduced birth weight. For term SGA, exposure to  $O_3$  during the 5th-8th gestational months or the 14th-17th and 21st-26th gestational weeks was associated with an increased risk of term SGA. Although not statistically significant, the association

was marginally significant during the 18th-20th gestational weeks for term SGA.

#### 4. Discussion

Leveraging this nationwide study of  $\sim 2.2$  million singleton term birth across 453 populous US counties, we estimated the association between maternal exposure to  $O_3$  and term birth weight and term SGA. We found that maternal exposure to  $O_3$  was associated with lower term birth weight and higher risk of term SGA. The potential critical exposure windows were the 13th- 25th and 32nd -37th gestational weeks for term birth weight and the 13th- 25th for term SGA.

We found that maternal exposure to  $O_3$  was associated with lower term birth weight or higher risk of term SGA, which is consistent with a systematic review summarizing six studies of  $O_3$  and term birth weight (Klepac et al. 2018). The pooled *meta*-analytic estimates of reduced birth weight associated with 10 ppb increase in  $O_3$  over the full pregnancy ranged from -4.6 g to -27.3 g (Klepac et al. 2018), which was comparable with our estimates of -7.6 g. Our findings are also in line with a recent multi-city study in China, which found a 5.7 g (95% CI: -10.1 g, -1.3 g) reduction in birth weight per 10 ppb increase in  $O_3$  over the full pregnancy (Guo et al. 2020). However, our findings are in contrast with a few previous studies, which found no evidence of any association between maternal exposure to  $O_3$  and lower rates of fetal growth (Chen et al. 2002; Gouveia et al. 2004; Stieb et al. 2012).

Although the biological plausibility for the association between  $\rm O_3$  exposure and fetal growth is not clear yet, oxidative stress and systemic inflammation may play a role (Bignami et al. 1994; Gunnison and Hatch 1999; Larini and Bocci 2005; Schoots et al. 2018; Thibodeaux et al. 2003). Fetal weight grows dramatically from the second trimester and fetal growth requires adequate oxygen and nutrient. Exposure to  $\rm O_3$  triggers the release of lipid peroxidation products and inflammatory cytokines into circulation (Larini and Bocci 2005). These oxidative and inflammatory products could harm placental circulation and function, leading to placental insufficiency and therefore affecting fetal growth (Larini and Bocci 2005). The  $\rm O_3$  exposure induced birth weight

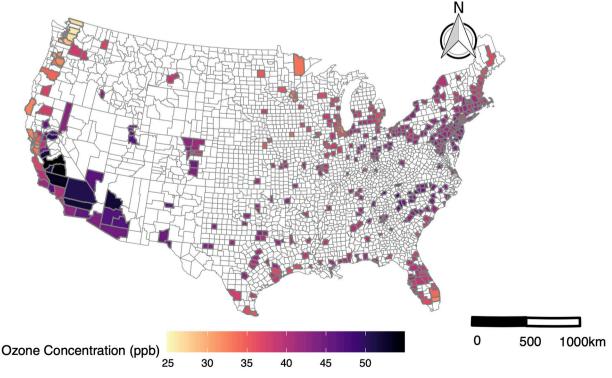


Fig. 1. Annual average ozone concentrations among 453 US counties in 2002.

<sup>&</sup>lt;sup>b</sup> Smoking and drinking were not recorded on California birth certificates and we assigned an unknown category of smoking and alcohol status for all California births.

<sup>&</sup>lt;sup>c</sup> Season of conception: spring (March-May), summer (June-August), fall (September-November), and winter (December-February).

 $<sup>^{\</sup>rm d}$  The percentage of people in a county below the federal poverty line. Abbreviations: SGA = small for gestational age; SD = standard deviation.

Table 2
Changes in birth weight (grams) and odds ratio of term small for gestational age associated with 10 parts per billion (ppb) increase in ozone exposure in the United States.

Model	Trimester	Term birth weight [change (g) (95 % CI)]	Term SGA [OR (95 % CI)]
Main model <sup>a</sup>	Entire	-7.6 (-8.8, -6.4)	1.030 (1.020,
	pregnancy		1.040)
	First trimester	-1.3 (-2.2, $-0.4$ )	1.005 (0.998,
			1.012)
	Second	-3.3 (-4.3, $-2.3$ )	1.016 (1.008,
	trimester		1.024)
	Third	-2.6 (-3.4, $-1.7$ )	1.006 (1.000,
	trimester		1.013)
$Main model + PM_{2.5}^{b}$	Entire	-7.3 (-8.6, -6.0)	1.030 (1.020,
	pregnancy		1.040)
	First trimester	-1.1 (-2.0, $-0.2$ )	1.005 (0.998,
			1.012)
	Second	-3.3 (-4.3, $-2.3$ )	1.016 (1.008,
	trimester		1.024)
	Third	-2.4 (-3.3, -1.6)	1.006 (1.000,
	trimester		1.013)
Main model -	Entire	-7.8 (-8.9, -6.6)	1.029 (1.020,
Temperature <sup>c</sup>	pregnancy		1.038)
	First trimester	-1.4 (-2.3, $-0.4$ )	1.005 (0.997,
			1.012)
	Second	-3.4 (-4.3, $-2.5$ )	1.014 (1.007,
	trimester		1.021)
	Third	-2.6 (-3.4, -1.7)	1.007 (1.001,
	trimester		1.014)

<sup>&</sup>lt;sup>a</sup> Models were adjusted for maternal age, maternal race/ethnicity, maternal educational attainment, marital status, chronic hypertension, infant sex, parity, alcohol drinking during pregnancy, smoking during pregnancy, timing of initiation of prenatal care, season of conception, mean ambient temperature during the entire pregnancy using natural cubic splines with 3 degrees of freedom, county poverty rate, census region, and mutually adjusted for ozone exposure in other trimesters. Term birth weight models were also adjusted for gestational length.

reduction was observed in animal studies (Bignami et al. 1994; Gunnison and Hatch 1999; Thibodeaux et al. 2003). Mid and late pregnancy exposure to O<sub>3</sub> reduced mice weight gain (Bignami et al. 1994; Thibodeaux et al. 2003).

We found that the association was stronger among non-Hispanic Black, unmarried mothers, or mothers with a lower school education, which could be due to those mothers being more likely to be exposed to a higher level of air pollution (Jbaily et al. 2022; Lamichhane et al. 2020). For example, in a descriptive analysis linking demographic data and PM<sub>2.5</sub> data among 32,000 US zip codes, the non-Hispanic population was exposed to  $\sim 14$  % higher level of PM<sub>2.5</sub> concentration than those of the white population in 2016 (Jbaily et al. 2022). The more pronounced associations among pregnant women with lower socioeconomic status might also due to a lack of awareness or access to maternal health care services (Lamichhane et al. 2020; Shah and Births 2010). Our finding of no heterogeneity between male and female infants is consistent with the previous study (Guo et al. 2020).

To identify critical exposure windows, most prior studies (Gouveia et al. 2004; Lamichhane et al. 2020; Olsson et al. 2020) ignore  $\rm O_3$  exposure in other exposure windows, which might lead to identify inaccurate critical windows (Wilson et al. 2017). To our knowledge, we only found two studies that modelled monthly or weekly  $\rm O_3$  exposure using distributed lag models to identify critical exposure windows for  $\rm O_3$  and term birth weight (Darrow et al. 2011; Wang et al. 2021). Our findings are in agreement with the Chinese study conducted in

**Table 3**Stratified analysis for changes in birth weight (grams) and odds ratio of term small for gestational age associated with 10 parts per billion (ppb) increase in ambient ozone during the entire pregnancy.

Variables	Term birth weight	Term SGA		
	Change in birth weight (g) (95 % CI)	P <sup>a</sup> <sub>Heterogeneity</sub>	OR	P <sup>a</sup> Heterogeneit
Maternal age (years)		< 0.001		0.353
18–24	-12.9 (-15.1,		1.046	
	-11.0)		(1.030,	
			1.063)	
25–29	-9.9 (-12.3,		1.031	
	-8.0)		(1.012,	
			1.051)	
30–34	-5.7 (-8.2,		1.020	
	-3.0)		(0.999,	
			1.042)	
35–39	-3.1 (-6.7, 1.0)		1.025	
			(0.995,	
			1.056)	
≥40	1.4 (-6.5, 9.0)		1.032	
			(0.972,	
3.6-4		-0.001	1.095)	-0.001
Maternal race/		< 0.001		< 0.001
ethnicity	0 5 ( 11 0		1 025	
Hispanic	-9.5 (-11.8, -7.0)		1.035	
	-7.0)		(1.017, 1.054)	
Non-Hispanic	-6.7 (-10.8,		1.045	
White	-3.0)		(1.019,	
***************************************	0.0)		1.072)	
Non-Hispanic	-13.8 (-15.6,		1.060	
Black	-12.0)		(1.043,	
			1.076)	
Other	8.1 (3.9, 12.0)		0.977	
			(0.948,	
			1.006)	
Educational attainment		< 0.001		< 0.001
Lower than high	-6.1 (-10.8,		1.030	
school	-1.0)		(0.994,	
			1.066)	
High school	-13.2 (-15.1,		1.055	
	-11.0)		(1.040,	
			1.070)	
Higher than	-5.1 (-6.8,		1.008	
high school	-3.0)		(0.993,	
Monital at-ti-		<0.001	1.023)	0.000
Marital status Married	62(70	< 0.001	1 021	0.006
iviai fieu	-6.3 (-7.8, -5.0)		1.021	
	~J.0)		(1.008, 1.034)	
Unmarried	-11.6 (-13.9,		1.054)	
Omnai i i cu	-11.0 (-13.9, -9.0)		(1.034,	
	2.0)		1.067)	
Infant sex		0.305	/	0.273
Male	-8.3 (-10.0,		1.025	
-	-6.0)		(1.012,	
	•		1.039)	
Female	-6.9 (-8.6,		1.036	
	-5.0)		(1.022,	
			1.051)	

Abbreviations: SGA = small for gestational age; OR = odds ratio.

Models were adjusted for maternal age, maternal race/ethnicity, maternal educational attainment, marital status, chronic hypertension, infant sex, parity, alcohol drinking during pregnancy, smoking during pregnancy, timing of initiation of prenatal care, season of conception, mean ambient temperature during the entire pregnancy using natural cubic splines with 3 degrees of freedom, county poverty rate, and census region. Term birth weight models were also adjusted for gestational length.

b Main models with additionally adjusted for PM<sub>2.5</sub>.

 $<sup>^{\</sup>rm c}$  Main models without adjustment for mean ambient temperature during the entire pregnancy using natural cubic splines with 3 degrees of freedom. Abbreviations: OR = odds ratio; CI = confidence interval; SGA = small for gestational age; PM<sub>2.5</sub> = fine particulate matter.

<sup>&</sup>lt;sup>a</sup>We used a Wald statistic to test the heterogeneity across subgroups.

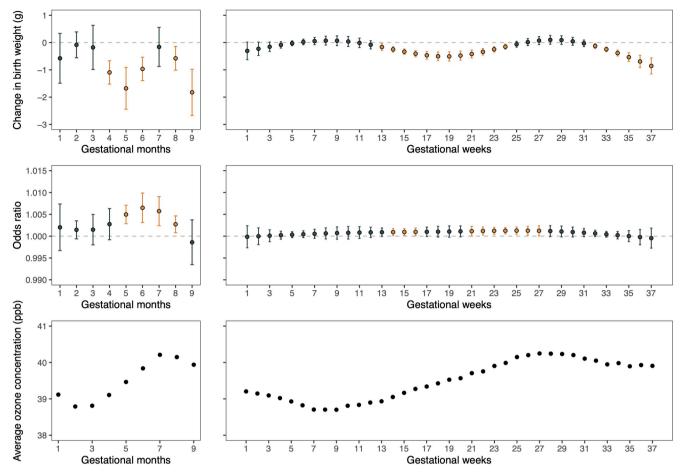


Fig. 2. Changes in birth weight (top panel) and odds ratio of term small for gestational age (middle panel) associated with 10 parts per billion (ppb) increase in monthly and weekly ozone exposure and the average ozone concentration across gestational months and weeks (bottom panel) during the pregnancy among term singleton births in the United States, 2002. Models were adjusted for maternal age, maternal race/ethnicity, maternal educational attainment, marital status, chronic hypertension, infant sex, parity, alcohol drinking during pregnancy, smoking during pregnancy, timing of initiation of prenatal care, season of conception, mean ambient temperature during the entire pregnancy using natural cubic splines with 3 degrees of freedom, county poverty rate, and census region. Term birth weight models were also adjusted for gestational length.

Guangzhou, which identified the critical exposure window for  $O_3$  and low birth weight during the 15th-26th gestational weeks (Wang et al. 2021)

Our study has several limitations. First, we used the populationweighted average of county-level O<sub>3</sub> exposure instead of personal O<sub>3</sub> exposure, which might lead to exposure misclassification. However, this potential exposure misclassification is more likely to be nondifferential and lead to underestimation of the associations (Kioumourtzoglou et al. 2014). Second, we did not consider residential mobility in our exposure assessment due to a lack of data, which might introduce another source of exposure misclassification. However, previous studies (Chen et al. 2010; Pennington et al. 2017) suggested that only a small proportion of pregnant women relocated during their pregnancies and much less moved to another county, suggesting that bias from this source of exposure misclassification should be limited. Third, compared to particulate matter, the spatial resolution of O3 is relatively low. The low spatial resolution of exposure estimates may result in increased bias and confounding (McIsaac et al. 2021), leading to underestimate the associations. Fourth, although we adjusted for maternal characteristics, infant sex, season of conception, county-level factors, and meteorological factors, residual confounding is still possible, such as traffic noise (Smith et al. 2017). Fifth, we included only 453 US counties with the availability of both O<sub>3</sub> and the births data, and our findings may not be generalizable to the remaining counties. We only used the 2002 birth data, and our results may not be generalizable to other years. Also, early

delivery mediates the effects of  $O_3$  on birth weight. To evaluate the direct effects of  $O_3$  on birth weight, we restricted to term births, and findings of our study cannot be generalizable to all births.

One strength of our study is the large sample size, including  $\sim 2.2$  million US singleton births across different geographic regressions in the US. Also, we used distributed lag models fitting monthly and weekly  $\rm O_3$  exposure during pregnancy to identify potential critical exposure windows. The consistency in identified critical windows using different temporal scales confirmed the robustness of our findings.

#### 5. Conclusions

In summary, among  $\sim 2.2$  million singleton live births across 453 US counties, we found that maternal exposure to  $O_3$  was associated with lower birth weight and a higher risk of small for gestational age, and the critical exposure windows were the 13th- 25th gestational weeks. Our findings might be useful to prevention of fetal growth restriction.

#### CRediT authorship contribution statement

**Shengzhi Sun:** Conceptualization, Writing – original draft, Formal analysis, Data curation, Investigation. **Jiajia Wang:** Formal analysis, Methodology, Writing – review & editing. **Wangnan Cao:** Conceptualization, Writing – original draft, Data curation, Investigation. **Lizhi Wu:** Writing – review & editing. **Yu Tian:** Writing – review & editing. **Feng** 

Sun: Writing – review & editing. Zhenyu Zhang: Writing – review & editing. Yang Ge: Writing – review & editing. Jianqiang Du: Writing – review & editing. Xiaobo Li: Writing – review & editing. Rui Chen: Conceptualization, Resources, Writing – review & editing.

#### **Declaration of Competing Interest**

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

#### Data availability

Data will be made available on request.

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#### Appendix A. Supplementary material

Supplementary data to this article can be found online at https://doi.org/10.1016/j.envint.2022.107554.

#### References

- Alexander, G.R., Himes, J.H., Kaufman, R.B., Mor, J., Kogan, M., 1996. A United States national reference for fetal growth. Obstet. Gynecol. 87, 163–168.
- Bekkar, B., Pacheco, S., Basu, R., DeNicola, N., 2020. Association of air pollution and heat exposure with preterm birth, low birth weight, and stillbirth in the US: a systematic review. JAMA Netw. Open 3, e208243.
- Berrocal, V.J., Gelfand, A.E., Holland, D.M., 2010. A bivariate space-time downscaler under space and time misalignment. Ann. Appl. Stat. 4, 1942.
- Berrocal, V.J., Gelfand, A.E., Holland, D.M., 2012. Space-time data fusion under error in computer model output: an application to modeling air quality. Biometrics 68, 837–848.
- Bignami, G., Musi, B., Dellomo, G., Laviola, G., Alleva, E., 1994. Limited effects of ozone exposure during pregnancy on physical and neurobehavioral development of CD-1 mice. Toxicol. Appl. Pharmacol. 129, 264–271.
- Buckley, J.P., Samet, J.M., Richardson, D.B., 2014. Commentary: Does air pollution confound studies of temperature? Epidemiology 25, 242–245.
- Byun, D.; Schere, K.L. Review of the governing equations, computational algorithms, and other components of the Models-3 Community Multiscale Air Quality (CMAQ) modeling system. 2006.
- Chen, L., Yang, W., Jennison, B.L., Goodrich, A., Omaye, S.T., 2002. Air pollution and birth weight in northern Nevada, 1991–1999. Inhal. Toxicol. 14, 141–157.
- Chen, L., Bell, E.M., Caton, A.R., Druschel, C.M., Lin, S., 2010. Residential mobility during pregnancy and the potential for ambient air pollution exposure misclassification. Environ. Res. 110, 162–168.
- Dai, Y., Qiu, H., Sun, S., Yang, Y., Lin, H., Tian, L., 2018. Age-dependent effect of ambient ozone on emergency asthma hospitalizations in Hong Kong. J. Allergy Clin. Immunol. 141 (1532–1534), e1535.
- Darrow, L.A., Klein, M., Strickland, M.J., Mulholland, J.A., Tolbert, P.E., 2011. Ambient air pollution and birth weight in full-term infants in Atlanta, 1994–2004. Environ. Health Perspect 119, 731–737.
- EPA, U. Integrated Science Assessment (ISA) for Ozone and Related Photochemical Oxidants. US Environmental Protection Agency Washington, DC, USA; 2020.
- Gouveia, N., Bremner, S., Novaes, H.M.D., 2004. Association between ambient air pollution and birth weight in S\(\tilde{a}\) Paulo. Brazil. J. Epidemiol. Community Health 58, 11–17
- Gunnison, A.F., Hatch, G.E., 1999.  $O_3$ -induced inflammation in prepregnant, pregnant, and lactating rats correlates with  $O_3$  dose estimated by 18O. Am. J. Physiol. Lung Cell. Mol. 276, L332–L340.
- Guo, P.; Chen, Y.; Wu, H.; Zeng, J.; Zeng, Z.; Li, W.; Zhang, Q.; Huo, X.; Feng, W.; Lin, J. Ambient air pollution and markers of fetal growth: A retrospective population-based cohort study of 2.57 million term singleton births in China. Environ. Int. 135, (2020). 105410.
- Horikoshi, M., Beaumont, R.N., Day, F.R., Warrington, N.M., Kooijman, M.N., Fernandez-Tajes, J., Feenstra, B., Van Zuydam, N.R., Gaulton, K.J., Grarup, N., 2016. Genome-wide associations for birth weight and correlations with adult disease. Nature 538, 248–252.
- Jbaily, A., Zhou, X., Liu, J., Lee, T.-H., Kamareddine, L., Verguet, S., Dominici, F., 2022. Air pollution exposure disparities across US population and income groups. Nature 601, 228–233.
- Kioumourtzoglou, M.-A., Spiegelman, D., Szpiro, A.A., Sheppard, L., Kaufman, J.D., Yanosky, J.D., Williams, R., Laden, F., Hong, B., Suh, H., 2014. Exposure

- measurement error in  $PM_{2.5}$  health effects studies: a pooled analysis of eight personal exposure validation studies. Environ. Health 13, 1–11.
- Klepac, P., Locatelli, I., Korošec, S., Künzli, N., Kukec, A., 2018. Ambient air pollution and pregnancy outcomes: A comprehensive review and identification of environmental public health challenges. Environ. Res. 167, 144–159.
- Lamichhane, D.K., Lee, S.-Y., Ahn, K., Kim, K.W., Shin, Y.H., Suh, D.I., Hong, S.-J., Kim, H.-C., 2020. Quantile regression analysis of the socioeconomic inequalities in air pollution and birth weight. Environ. Int. 142, 105875.
- Larini, A., Bocci, V., 2005. Effects of ozone on isolated peripheral blood mononuclear cells. Toxicol. In Vitro 19, 55–61.
- Li, X., Huang, S., Jiao, A., Yang, X., Yun, J., Wang, Y., Xue, X., Chu, Y., Liu, F., Liu, Y., 2017. Association between ambient fine particulate matter and preterm birth or term low birth weight: an updated systematic review and meta-analysis. Environ. Pollut. 227, 596–605.
- Malhotra, A., Allison, B.J., Castillo-Melendez, M., Jenkin, G., Polglase, G.R., Miller, S.L., 2019. Neonatal morbidities of fetal growth restriction: pathophysiology and impact. Front. Endocrinol. 10, 55.
- McIsaac, M.A., Sanders, E., Kuester, T., Aronson, K.J., Kyba, C.C., 2021. The impact of image resolution on power, bias, and confounding: A simulation study of ambient light at night exposure. Environ. Epidemiol. 5, e145.
- Mirabelli, M.C., Vaidyanathan, A., Flanders, W.D., Qin, X., Garbe, P., 2016. Outdoor PM<sub>2.5</sub>, ambient air temperature, and asthma symptoms in the past 14 days among adults with active asthma. Environ. Health Perspect. 124, 1882–1890.
- Morello-Frosch, R., Jesdale, B.M., Sadd, J.L., Pastor, M., 2010. Ambient air pollution exposure and full-term birth weight in California. Environ. Health 9, 1–13.

#### NCHS. Vital Statistics Online Data Portal. 2022.

- Niu, Z., Zhang, J., Chu, L., Wang, M., Guo, Y., Qiao, L., Farhat, Z., Grippo, A., Xu, J., Mu, L., 2020. Ambient ozone exposure during preconception and gestation periods and birth weight: when is the critical window? ISEE Conference Abstracts.
- Niu, Y., Zhou, Y., Chen, R., Yin, P., Meng, X., Wang, W., Liu, C., Ji, J.S., Qiu, Y., Kan, H., 2022. Long-term exposure to ozone and cardiovascular mortality in China: a nationwide cohort study. Lancet Planet Health 6, e496–e503.
- Oken, E., Kleinman, K.P., Rich-Edwards, J., Gillman, M.W., 2003. A nearly continuous measure of birth weight for gestational age using a United States national reference. BMC Pediatr. 3, 1–10.
- Olsson, D., Johansson, C., Forsberg, B., 2020. Associations between vehicle exhaust particles and ozone at home address and birth weight. Int. J. Environ. Res. Public Health 17, 3836.
- Pennington, A.F., Strickland, M.J., Klein, M., Zhai, X., Russell, A.G., Hansen, C., Darrow, L.A., 2017. Measurement error in mobile source air pollution exposure estimates due to residential mobility during pregnancy. J. Expo Sci. Environ. Epidemiol. 27, 513–520.
- Rappazzo, K.M., Nichols, J.L., Rice, R.B., Luben, T.J., 2021. Ozone exposure during early pregnancy and preterm birth: A systematic review and meta-analysis. Environ. Res. 198. 111317.
- Rothman, K.J.; Greenland, S.; Lash, T.L. Modern epidemiology ed^eds: Wolters Kluwer Health/Lippincott Williams & Wilkins Philadelphia; (2008).
- Schoots, M.H., Gordijn, S.J., Scherjon, S.A., van Goor, H., Hillebrands, J.-L., 2018. Oxidative stress in placental pathology. Placenta 69, 153–161.
- Shah, P.S.; Births, K.S.G.o.D.o.P.L.B. Paternal factors and low birthweight, preterm, and small for gestational age births: a systematic review. Am. J. Obstet. Gynecol. 202 (2010), 103–123.
- Smith, R.B., Fecht, D., Gulliver, J., Beevers, S.D., Dajnak, D., Blangiardo, M., Ghosh, R.E., Hansell, A.L., Kelly, F.J., Anderson, H.R., 2017. Impact of London's road traffic air and noise pollution on birth weight: retrospective population based cohort study. BMJ 359, j5299.
- Spangler, K.R., Weinberger, K.R., Wellenius, G.A., 2019. Suitability of gridded climate datasets for use in environmental epidemiology. J. Expo Sci. Environ. Epidemiol. 29, 777–789.
- Stieb, D.M., Chen, L., Eshoul, M., Judek, S., 2012. Ambient air pollution, birth weight and preterm birth: a systematic review and meta-analysis. Environ. Res. 117, 100–111.
- Strosnider, H.M., Chang, H.H., Darrow, L.A., Liu, Y., Vaidyanathan, A., Strickland, M.J., 2019. Age-specific associations of ozone and fine particulate matter with respiratory emergency department visits in the United States. Am. J. Respir. Crit. Care Med. 199, 882–890.
- Sun, X., Luo, X., Zhao, C., Zhang, B., Tao, J., Yang, Z., Ma, W., Liu, T., 2016. The associations between birth weight and exposure to fine particulate matter (PM<sub>2.5</sub>) and its chemical constituents during pregnancy: A meta-analysis. Environ. Pollut. 211, 38–47.
- Sun, S., Spangler, K.R., Weinberger, K.R., Yanosky, J.D., Braun, J.M., Wellenius, G.A., 2019a. Ambient temperature and markers of fetal growth: a retrospective observational study of 29 million US singleton births. Environ. Health Perspect. 127, 067005.
- Sun, S., Weinberger, K.R., Spangler, K.R., Eliot, M.N., Braun, J.M., Wellenius, G.A., 2019b. Ambient temperature and preterm birth: a retrospective study of 32 million US singleton births. Environ. Int. 126, 7–13.
- Sun, S., Weinberger, K.R., Nori-Sarma, A., Spangler, K.R., Sun, Y., Dominici, F., Wellenius, G.A., 2021. Ambient heat and risks of emergency department visits among adults in the United States: time stratified case crossover study. BMJ 375, e065653.
- Thibodeaux, J.R., Hanson, R.G., Rogers, J.M., Grey, B.E., Barbee, B.D., Richards, J.H., Butenhoff, J.L., Stevenson, L.A., Lau, C., 2003. Exposure to perfluorooctane sulfonate during pregnancy in rat and mouse. I: maternal and prenatal evaluations. Toxicol. Sci. 74, 369–381.
- U.S. Census Bureau. Small Area Income and Poverty Estimates. 2022.

- Wang, Q., Miao, H., Warren, J.L., Ren, M., Benmarhnia, T., Knibbs, L.D., Zhang, H., Zhao, Q., Huang, C., 2021. Association of maternal ozone exposure with term low birth weight and susceptible window identification. Environ. Int. 146, 106208.
- birth weight and susceptible window identification. Environ. Int. 146, 106208. Wilson, A., Chiu, Y.-H.-M., Hsu, H.-H.-L., Wright, R.O., Wright, R.J., Coull, B.A., 2017. Potential for bias when estimating critical windows for air pollution in children's health. Am. J. Epidemiol. 186, 1281–1289.
- Zhang, Z., Kris-Etherton, P.M., Hartman, T.J., 2014. Birth weight and risk factors for cardiovascular disease and type 2 diabetes in US children and adolescents: 10 year results from NHANES. Matern. Child Health J. 18, 1423–1432.