The Combined Effects of Fine Particulate Matter and Temperature on Preterm Birth in Seoul, 2010–2016

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Abstract: Background: Preterm birth contributes to the morbidity and mortality of newborns and infants. Recent studies have shown that maternal exposure to particulate matter and extreme temperatures results in immune dysfunction, which can induce preterm birth. This study aimed to evaluate the association between fine particulate matter (PM$_{2.5}$) exposure, temperature, and preterm birth in Seoul, Republic of Korea. Methods: We used 2010–2016 birth data from Seoul, obtained from the Korea National Statistical Office Microdata. PM$_{2.5}$ concentration data from Seoul were generated through the Community Multiscale Air Quality (CMAQ) model. Seoul temperature data were collected from the Korea Meteorological Administration (KMA). The exposure period of PM$_{2.5}$ and temperature were divided into the first (TR1), second (TR2), and third (TR3) trimesters of pregnancy. The mean PM$_{2.5}$ concentration was used in units of $\times 10 \, \mu g/m^3$ and the mean temperature was divided into four categories based on quartiles. Logistic regression analyses were performed to evaluate the association between PM$_{2.5}$ exposure and preterm birth, as well as the combined effects of PM$_{2.5}$ exposure and temperature on preterm birth. Result: In a model that includes three trimesters of PM$_{2.5}$ and temperature data as exposures, which assumes an interaction between PM$_{2.5}$ and temperature in each trimester, the risk of preterm birth was positively associated with TR1 PM$_{2.5}$ exposure among pregnant women exposed to relatively low mean temperatures ($<3.4 \, ^oC$) during TR1 (OR 1.134, 95% CI 1.061–1.213, $p < 0.001$). Conclusions: When we assumed the interaction between PM$_{2.5}$ exposure and temperature exposure, PM$_{2.5}$ exposure during TR1 increased the risk of preterm birth among pregnant women exposed to low temperatures during TR1. Pregnant women should be aware of the risk associated with combined exposure to particulate matter and low temperatures during TR1 to prevent preterm birth.

Keywords: preterm birth; low birth weight; PM$_{2.5}$; temperature

1. Introduction

Preterm birth is defined as labor and delivery at less than 37 gestational weeks. Preterm birth is associated with stillbirths and premature infant complications, resulting in substantial economic costs and impairments that persist into adulthood [1]. The high
risk of morbidity and mortality among premature infants is partly attributable to growth restrictions associated with premature birth. In the long term, these conditions increase the risk of a variety of health hazards [2]. In Korea, birth outcomes among women of childbearing age are generally deteriorating [3]. This trend is maintained regardless throughout the childbearing years among women in Korea, where the low birth rate problem is serious [3].

According to a World Health Organization (WHO) report, maternal exposure to air pollution increases risk of preterm birth and low birth weight (LBW) (WHO, 2018). Fine particulate matter (PM$_{2.5}$, or particles with an aerodynamic diameter of 2.5 µm or less) is more uniformly distributed in air compared with other pollutants. PM$_{2.5}$ exposure is a useful indicator of overall exposure to outdoor particulate matter (PM) and can be used to study the health effects of such exposure [4]. Exposure to PM$_{2.5}$ is strongly associated with adverse health effects [5]. Many studies have suggested that exposure to PM above a certain concentration affects the immune system [6]. PM$_{2.5}$ enters the body and accumulates in various compartments of the respiratory tract, interacting with epithelial and immune cells [7]. Because the immune system is the most sensitive target of PM$_{2.5}$, it is important to elucidate the impact of air pollutants on human immune function [8]. Inhaled PM$_{2.5}$ can deposit in other compartments of the respiratory tract and interact with epithelial and immune cells. Consequentially, exposure to PM$_{2.5}$ can cause local or systemic inflammatory responses [7].

Previous studies have reported that impaired immunity among pregnant women is associated with preterm birth [9,10]. Particulate matter and extreme temperatures increase risk of preterm birth by facilitating oxidative stress and interrupting the immune system. It has been suggested that inflammation and infection around the placenta are the potential biological mechanism underlying the association between air pollution and preterm birth [10].

The recent increase in environmental pollution, which is associated with high PM concentrations, has been shown to have negative effects on maternal immunity and neonatal health [11]. PM exposure may induce reproductive toxicity mediated by placental DNA methylation and maternal inflammatory responses, suggesting a link between PM exposure and preterm birth risk [12].

Changes in environmental temperatures may also affect maternal immune system [13]. Further, simultaneous exposure to PM and extreme temperatures during pregnancy may adversely impact immune function among pregnant women, which may trigger pregnancy complications [9,10]. For instance, several studies have reported that exposure to both PM and environmental temperature changes affect fetal growth and increase risk of premature birth or LBW [14,15]. However, few studies have investigated the combined effects of exposure to PM$_{2.5}$ and ambient temperature on risk of preterm birth. To address this gap in the literature, the present study investigated the effect of prenatal exposure to PM$_{2.5}$ in each trimester, as well as the combined effect of maternal PM$_{2.5}$ exposure and temperature, on risk of preterm birth.

2. Materials and Methods

2.1. Birth Data

We used birth data from the MicroData Integrated Service (MDIS) system of the Korea National Statistical Office (https://mdis.kostat.go.kr). The MDIS system is a national statistical site containing sectorwide household and population data, including birth outcomes in the database.

A total of 605,026 babies were born from January 2010 through December 2016 in Seoul, Republic of Korea. After excluding the infants born (i) before 22 weeks or after 42 weeks of gestation and (ii) as part of multiple pregnancies (e.g., twins, triplets), 581,239 infants were included in our main analysis.
We used data from birth surveys, which included data on sex, birth weight, birth date, gestational age at birth, parity, season of birth, parental age, parental education status, parental occupations, and residential area.

2.2. Exposure Assessment

PM$_{2.5}$ levels were measured using the Community Multiscale Air Quality (CMAQ) Modeling System [4]. Between 2009 and 2016, PM$_{2.5}$ concentrations were estimated using the Weather Research and Forecast (WRF, version 3.3.1)—Sparse Matrix Operator Kernel Emission (SMOKE, version 3.1) Community Multiscale Air Quality (CMAQ, version 4.7.1) Modeling System [16]. We used the CMAQ data as PM$_{2.5}$ exposure data because PM$_{2.5}$ monitoring data has many missing values.

The CMAQ modeling data were generated using meteorological research and forecasting models comprising three overlapping weather data sources at 27, 9, and 3 km for a specific period of time [17]. The data are more representative of meteorological and atmospheric conditions than the Meteorological Agency’s monitoring data [18]. Additionally, the hourly CMAQ data have been collected at the administrative district level without any missing values. Daily data were calculated by averaging hourly CMAQ data over a 24-h period.

The temperature data from 2009 through 2016 were obtained from the Korea Meteorological Administration (KMA, www.kma.go.kr). The daily mean temperatures of each TR were divided into the following groups according to the quartile values: group 1 (−14.5 to 3.4 °C), group 2 (3.4–14.5 °C), group 3 (14.5–22.8 °C), and group 4 (22.8–31.8 °C).

2.3. Statistical Analysis

For the comparison between full-term and preterm babies, chi-square test was used for categorical variables, and Student’s $t$-test for continuous variables. Categorical variables are described as frequencies and percentages, and continuous variables are reported as means and standard deviations.

The mean PM$_{2.5}$ and temperature exposure levels were calculated for each gestation period. During the prenatal period, each date was inversely estimated using gestational weeks provided in the microdata. PM$_{2.5}$ data, temperature data, and birth data were integrated and analyzed using common local unit addresses and dates. The PM and temperature exposures were merged using the calculated dates and addresses, and the prenatal exposure periods were divided into trimesters based on the birth data and gestational age at birth to analyze the impact of exposure by period [19]: 1 to 12 weeks in the first trimester (TR1), 13 to 26 weeks in the second TR (TR2), and 27 to 40 weeks in the third TR (TR3).

In the analysis, the mean PM$_{2.5}$ concentration was used in units of $\times 10 \mu g/m^3$ and temperature was analyzed in terms of quartiles: range 1 (−14.5 to 3.4 °C), range 2 (3.4–14.5 °C), range 3 (14.5–22.8 °C), and range 4 (22.8–31.8 °C).

Logistic regression models were used to evaluate the associations of prenatal exposure to PM$_{2.5}$ and extreme temperature with risk of preterm birth. Model 1 included PM$_{2.5}$ and temperature exposures of each trimester. Model 2 included PM$_{2.5}$ and temperature exposures of all three trimesters and assumed the interactions between PM$_{2.5}$ and temperature in each trimester. All models included child sex, mother’s age, mother’s educational level, parity, and season of birth as covariates. The models were calculated as follows:

Model 1: $\text{logit}(\pi) = \beta_0 + \beta_{\text{TR}i} + \text{covariates}$. TR$_i$: PM$_{2.5}$ in the ith trimester, $i = 1, 2, 3$.

Model 2: $\text{logit}(\pi) = \beta_0 + \beta_{\text{TR}1} + \beta_{\text{TR}2} + \beta_{\text{TR}3} + \text{covariates}$. TR1: PM$_{2.5}$ in the first trimester. TR2: PM$_{2.5}$ in the second trimester. TR3: PM$_{2.5}$ in the third trimester.

SAS version 9.4 (SAS Institute Inc., Cary, NC, USA) was used to perform all statistical analyses. The significance level was set at $p < 0.05$. 
3. Results

The characteristics of the full-term and preterm births are presented in Table 1. Higher maternal age and lower maternal education were associated with higher preterm birth rates. Table 2 presents the quartiles of temperature and PM$_{2.5}$ from March 2009 through February 2016 in Seoul.

Table 1. Baseline characteristics of full-term and premature babies.

| Characteristic                        | Full-Term (GA ≥ 37 Weeks) | Premature (GA < 37 Weeks) | p     |
|---------------------------------------|---------------------------|---------------------------|-------|
|                                       | n (%) or Mean ± SD        |                           |       |
| Sex                                   |                           |                           |       |
| Male                                  | 283,774 (51.09)           | 14,616 (56.56)            | <0.0001|
| Female                                | 271,625 (48.91)           | 11,224 (43.44)            |       |
| Birth weight (kg)                     | 3.28 ± 0.38               | 2.43 ± 0.63               | <0.0001|
| Gestational age at birth (weeks)      |                           |                           |       |
| 1st                                   | 324,888 (58.61)           | 14,488 (56.67)            | <0.0001|
| 2nd                                   | 193,259 (34.86)           | 9033 (35.33)              |       |
| ≥3rd                                  | 36,164 (6.52)             | 2043 (7.99)               |       |
| Parity of mother                      |                           |                           |       |
| 1st                                   | 324,888 (58.61)           | 14,488 (56.67)            | <0.0001|
| 2nd                                   | 193,259 (34.86)           | 9033 (35.33)              |       |
| ≥3rd                                  | 36,164 (6.52)             | 2043 (7.99)               |       |
| Season of birth                       |                           |                           |       |
| Spring                                | 142,981 (25.74)           | 6300 (24.38)              | <0.0001|
| Summer                                | 135,175 (24.34)           | 6740 (26.08)              |       |
| Fall                                  | 137,980 (24.84)           | 6298 (24.37)              |       |
| Winter                                | 139,263 (25.07)           | 6502 (25.16)              |       |
| Age of mother (years)                 |                           |                           |       |
| None                                  | 242 (0.04)                | 17 (0.07)                 | <0.0001|
| Elementary                            | 811 (0.15)                | 61 (0.24)                 |       |
| Middle                                | 4646 (0.84)               | 315 (1.23)                |       |
| High                                  | 96,978 (17.53)            | 5636 (21.98)              |       |
| University                            | 386,257 (69.82)           | 16,756 (65.34)            |       |
| Graduate                              | 64,266 (11.62)            | 2860 (11.15)              |       |
| Employment status of mother           |                           |                           |       |
| No                                    | 295,870 (53.27)           | 14,393 (55.70)            | <0.0001|
| Yes                                   | 259,529 (46.73)           | 11,447 (44.30)            |       |
| Age of father (years)                 |                           |                           |       |
| None                                  | 34.38 ± 4.27              | 34.82 ± 4.57              | <0.0001|
| Elementary                            | 788 (0.14)                | 56 (0.22)                 | <0.0001|
| Middle                                | 4294 (0.78)               | 252 (1.01)                |       |
| High                                  | 94,446 (17.20)            | 5194 (20.75)              |       |
| University                            | 367,083 (69.82)           | 16,037 (64.08)            |       |
| Graduate                              | 82,398 (15.00)            | 3477 (13.89)              |       |
| Employment status of father           |                           |                           |       |
| No                                    | 18,164 (3.27)             | 905 (3.50)                | 0.0408 |
| Yes                                   | 537,235 (96.73)           | 24,935 (96.50)            |       |

Abbreviation: GA, gestational age; SD, standard deviation.

Table 2. Quartiles of daily mean temperature (°C) and PM$_{2.5}$ (µg/m$^3$) in Seoul between 2009 and 2016.

|                   | Min   | Q1    | Median | Q3     | Max   |
|-------------------|-------|-------|--------|--------|-------|
| Temperature (°C)   | −14.5 | 3.4   | 14.5   | 22.8   | 31.8  |
| PM$_{2.5}$ (µg/m$^3$) | 3.5   | 18.5  | 27.2   | 37.9   | 200   |

Abbreviation: Min, minimum; Q1, first quartile; Q3, third quartile; Max, maximum.

The risk of preterm births by PM$_{2.5}$ (10 µg/m$^3$) exposure in each trimester is presented in Table 3. Model 1 does not assume the interaction between temperature and PM, and did not include exposure variables of other trimesters. The odds ratio of TR2 and TR3 in Model 2 indicates that an increase in PM exposure is associated with a decrease in the risk of preterm birth. However, in Model 2, which assumes the interaction and includes exposure variables of other trimesters, the maternal PM exposure in the TR1 only is significantly associated with an increased risk of preterm birth.
### Table 3. Risk of preterm births by PM$_{2.5}$ (10 µg/m$^3$) exposure in three trimesters.

| Stage | Model 1 $^a$ | | Model 2 $^b$ | | 
|-------|--------------|-------|--------------|-------|
|       | OR 95% CI    | $p$   | OR 95% CI    | $p$   |
| TR1   | 0.965 0.920–1.012 0.143 |     | 1.134 1.061–1.213 <0.001 |     |
| TR2   | 0.907 0.861–0.955 <0.001 |     | 1.024 0.959–1.094 0.474 |     |
| TR3   | 0.947 0.902–0.993 0.025 |     | 1.023 0.968–1.081 0.426 |     |

Logistic regression models were used to calculate odds ratio (OR), 95% confidence interval (95% CI), and corresponding $p$-value. $^a$ Model 1 included PM$_{2.5}$ and temperature exposures of each trimester. The model adjusted for sex, mother’s age, mother’s educational level, parity, and birth season. $^b$ Model 2 included PM$_{2.5}$ and temperature exposures of all three trimesters and assumed the interactions between PM$_{2.5}$ and temperature in each trimester. The model adjusted for sex, mother’s age, mother’s educational level, parity, and birth season.

The risk of preterm birth due to the combined effect of extreme temperature and PM$_{2.5}$ exposure is shown in Figure 1. In the model that mutually adjusted for each trimester, risk of preterm birth was highest during TR1 when the mean temperature during TR1 was less than $3.4 \, ^\circ\text{C}$ (OR 1.134, 95% CI 1.061–1.213, $p < 0.001$). On the other hand, the risk of preterm birth was lowest when the temperature was $14.5–22.8 \, ^\circ\text{C}$ during TR1.

![Figure 1](image-url). The combined effect of temperature ($^\circ\text{C}$) and PM$_{2.5}$ ($\times10 \, \mu\text{g/m}^3$) in each trimester on the risk of preterm birth. These results are based on the adjusted Model 2, including PM$_{2.5}$ and temperature exposures of all three trimesters, and this results assume the interactions between PM$_{2.5}$ and temperature in each trimester. The model adjusted for sex, mother’s age, mother’s educational level, parity, and birth season.

### 4. Discussion

We found that the risk of preterm birth was associated with combined exposure to PM$_{2.5}$ and low temperatures during TR1.

According to a recent study, the risk of spontaneous preterm delivery increased following exposures to temperatures between 0 and $-6 \, ^\circ\text{C}$ [20]. Temperature can have
varying effects on adaptive immunity at multiple levels, with elevated temperatures inducing immune cell activation, function, and delivery, while reduced temperatures inhibit these processes [21]. Another study reported that exposure to air pollution at low temperatures induces complications associated with pregnancy [22]. The findings from these studies are consistent with our results. The impact of environmental extremes on the prevalence of preterm birth has been recognized [23]. Our results demonstrate that maternal prenatal exposure to extreme external temperatures can affect the timing of parturition.

Inhalation of PM$_{2.5}$ particles induces oxidative stress and inflammation in pregnant women, which can contribute to the onset of preterm birth [24]. Air pollutants are complex substances that increase the risk of premature birth or LBW via inflammation, oxidative stress, endocrine disruption, and oxygen transportation disorders across the placenta [25]. As a result, exposure to air pollutants, such as PM$_{2.5}$ during pregnancy, has also been associated with poor fetal growth [26]. Our results were consistent with the previous literature that TRI is an important period of fetal development characterized by cell proliferation, invasion, and differentiation [27]. Regulation of immune responses at the placental interface occurs during early pregnancy; dysregulation during this time is the main driver of premature labor [28].

In our study, exposures between 22.8 and 31.8 $^\circ$C were relatively lower than the range of high temperatures classically associated with premature birth. The high temperature range in Korea is 22.8–31.8 $^\circ$C. The U-shaped dose–response relationship was observed between ambient temperature and risk of preterm birth in this study. Our results for high temperature exposure may be explained as follows. First, temperature represents the value of the average exposure temperature. Therefore, temperature exposure might not be constantly high, and the temperature may be lower than the data values presented. Second, the high temperature range consists of measurements higher than 30 $^\circ$C. Third, in obstetrics and gynecology, “maintaining normothermia” refers to the comfortable and healthy environmental temperature ranging between 23 and 25 $^\circ$C, which is associated with a lower risk of fetal death due to hypothermia. In our study, the range of 22–31 $^\circ$C was relatively warm. In our study, if the temperature range was relatively narrow and the average exposure temperature was between 30 and 40 degrees, our results may be different.

It is important to explain not only the effect on premature birth of high temperatures but also the influence on premature birth of low temperatures. Many previous studies have demonstrated a U-shaped relationship between temperature and birth outcomes, reflecting a higher risk of premature births at extreme temperatures and a lower risk of premature births at moderate temperatures. Some relevant studies have focused on high temperatures above 40 $^\circ$C. Our study’s main observation was that maternal exposure to low ambient temperatures during pregnancy could influence the timing of parturition. Additionally, we showed a protective effect of moderate ambient temperatures. Other studies have also reported that low temperatures increased the risk of preterm birth as much as high temperatures. Moderately warm temperatures have also been shown to be helpful for both the pregnant woman and her fetus [29], which is consistent with our findings.

Our study accounted for both environmental epidemiological considerations and gynecological considerations. We also considered extreme temperatures on both ends of the scale. We investigated about 600,000 pregnancies from 2009 through 2016 using national data. This was not a cohort study observing detailed factors, but it was a study that comprehensively observed an association between human health and specific environmental determinants using national statistical data.

Since this was a retrospective study using national data, one limitation was that we were unable to control some factors that might have affected outcomes concurrently with the variables of interest. However, with this study of about 600,000 pregnant women from 2009 through 2016, we were able to comprehensively observe an association between human health and specific environmental determinants. Our study adjusted for sex, maternal age, maternal education level, parity, birth season, and temperature, but could
not adjust for exposure in the working environment and outdoor activity time. Another limitation was related to the context of exposure assessment. Indoor temperatures and indoor PM$_{2.5}$ exposure can be vastly different from their outdoor counterparts; the use of air conditioners and air purifiers indoors, for example, can contribute to such differences. These uncertainties about exposure settings should be considered when interpreting the study results.

5. Conclusions

We found that pregnant women exposed to low temperatures and PM$_{2.5}$ during TR1 had a significantly increased risk of preterm delivery. This indicates that pregnant women should avoid very low temperatures and exposure to high concentrations of air pollutants simultaneously, particularly during TR1. Appropriate temperature control and fine particle management during TR1 could be an important measure to prevent preterm births. We need to consider not only extremely high temperatures but also the impact on premature births of extremely low temperatures.

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