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Geospatial Association between Low Birth Weight and Arsenic in Groundwater in New Hampshire, USA

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Abstract

Background—There is increasing evidence of the role of arsenic in the etiology of adverse human reproductive outcomes. Since drinking water can be a major source of arsenic to pregnant women, the effect of arsenic exposure through drinking water on human birth may be revealed by a geospatial association between arsenic concentration in groundwater and birth problems, particularly in a region where private wells substantially account for water supply, like New Hampshire, US.

Methods—We calculated town-level rates of preterm birth and term low birth weight (term LBW) for New Hampshire, using data for 1997-2009 and stratified by maternal age. We smoothed the rates using a locally-weighted averaging method to increase the statistical stability. The town-level groundwater arsenic values are from three GIS data layers generated by the US Geological Survey: probability of local groundwater arsenic concentration > 1 µg/L, probability > 5 µg/L, and probability > 10 µg/L. We calculated Pearson's correlation coefficients (r) between the reproductive outcomes (preterm birth and term LBW) and the arsenic values, at both state and county levels.

Results—For preterm birth, younger mothers (maternal age < 20) have a statewide $r = 0.70$ between the rates smoothed with a threshold = 2,000 births and the town mean arsenic level based on the data of probability > 10 µg/L; For older mothers, $r = 0.19$ when the smoothing threshold = 3,500; A majority of county level r values are positive based on the arsenic data of probability > 10 µg/L. For term LBW, younger mothers (maternal age < 25) have a statewide $r = 0.44$ between the rates smoothed with a threshold = 3,500 and town minimum arsenic level based on the data of probability > 1 µg/L; For older mothers, $r = 0.14$ when the rates are smoothed with a threshold =

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1,000 births and also adjusted by town median household income in 1999, and the arsenic values are the town minimum based on probability $> 10 \mu\text{g/L}$. At the county level, for younger mothers positive r values prevail, but for older mothers it is a mix. For both birth problems, the several most populous counties - with 60-80% of the state's population and clustering at the southwest corner of the state - are largely consistent in having a positive r across different smoothing thresholds.

Conclusion—We found evident spatial associations between the two adverse human reproductive outcomes and groundwater arsenic in New Hampshire, US. However, the degree of associations and their sensitivity to different representations of arsenic level are variable. Generally, preterm birth has a stronger spatial association with groundwater arsenic than term LBW, suggesting an inconsistency in the impact of arsenic on the two reproductive outcomes. For both outcomes, younger maternal age has stronger spatial associations with groundwater arsenic.

Keywords

preterm birth; low birth weight; arsenic; groundwater; locally-weighted averaging smoothing; New Hampshire

Introduction

Infant mortality can be impacted by adverse reproductive outcomes, and these outcomes can be sensitive to many environmental influences [1-4]. Geospatial analysis has been used to study variation in the occurrence of the adverse outcomes [3-8] and its possible association with environmental factors, including point-source pollution, such as hazardous waste sites and toxic release sites [3,4], pesticide exposure [9], and air pollution [10-12].

In the past 15 years, epidemiologists have been particularly interested in the role of arsenic in the etiology of adverse human reproductive outcomes, especially arsenic exposure from drinking water sourced from groundwater. A geographic concentration of the research is Bangladesh [13-21]. At the individual level, some studies in Bangladesh provide evidence that arsenic exposure is associated with adverse reproductive outcomes. For example, one study in the Matlab region identified a significant association in 1,578 mother-infant pairs over the range of urinary arsenic concentrations of $<100 \mu\text{g/L}$ (but not over the entire range measured (6-978 $\mu\text{g/L}$)). Over the lower range, a 1.68g reduction in birth weight was seen for every 1 $\mu\text{g/L}$ increase in urinary arsenic concentration [18]. Another study in the Sirajdikhan region analyzed hair, toenail, and drinking water samples collected from 52 pregnant women at multiple time points during pregnancy and from their newborns after birth, and suggests that maternal arsenic exposure early in pregnancy negatively affects newborn birth weight [17]. However, one study that has examined 2,006 pregnant women from the Faridpur Sadar, Matlab, and Shahrasti regions who had chronically exposed to a range of naturally-occurring concentrations of arsenic in drinking-water only finds small (although statistically significant) association between arsenic exposure and birth-defects, and did not see such an association in some other outcomes [16]. At the ecological level, a study that examined fetal loss and infant death in the Matlab region performed geospatial clustering analysis on both reproductive outcomes and arsenic concentration. It finds that the spatial patterns of arsenic concentrations in tube-well water are linked with the adverse

pregnancy outcome clusters [20]. Besides Bangladesh, an individual-level study in India finds that exposure to high concentrations of arsenic (200 µg/L) during pregnancy was associated with a six-fold increased risk of stillbirth after adjustment for potential confounders, but finds no association between arsenic exposure and spontaneous abortion or overall infant mortality [22]. Ecologic studies in Taiwan and Chile indicate that arsenic endemic areas with drinking water contamination have significantly lower average birth weights compared to non-endemic regions [23,24]; however, an ecological study in Mongolia did not support an association [25]. The mechanism through which arsenic influences birth weight is not clear; one of many possible explanations is arsenic-induced impaired glucose tolerance during pregnancy [26,27].

This paper presents a geospatial analysis of associations between groundwater arsenic level and two adverse reproductive outcomes, including preterm birth and term LBW, in New Hampshire, US. The novelties of this study include: 1) to our knowledge, geographically this is one of the earliest studies of its kind particularly about a US cohort; 2) this might be one of few studies exploring effect of low level of groundwater arsenic on reproductive outcomes; and 3) methodologically, different from most ecological studies of its kind that compare two or a few selected regions, we compares the continuous geographic distributions of arsenic and adverse outcomes. Our conceptual model holds that, if indeed the arsenic in daily drinking water has an effect on human birth, this effect may be revealed by a correlation between the spatial variability of arsenic level in groundwater and the spatial variability of adverse reproductive outcomes, and that this may be particularly detectable in a region where private wells substantially account for the water supply. In New Hampshire, about 40% of the population uses private wells as a primary source for drinking water, which is a reason for us to choose it as our study area.

Data

Our choice of preterm birth (gestational period < 37 weeks) and LBW (birth weight < 2500g, e.g., [1]) for this study is first based on the availability of data, and is also following suggestions by Wilcox [1], who provides a powerful discussion of the importance of separating preterm birth and term birth in epidemiological studies. He states that “(a)n exposure that affects fetal growth does not necessarily affect the risk of preterm delivery”, and “(c)onversely, a factor that increases the risk of preterm delivery would not necessarily change the average weight of babies delivered at term”. He then recommends that when the data of gestational age are available, the preterm birth rate be selected for analysis. For term births, Wilcox recommends to use mean birth weight and simultaneously consider standard deviation (SD), which has been adopted by most studies that are comparing two or a few regions [e.g.,16,23-25]. However, simultaneously considering mean and SD is difficult when working with many areal units. In such a situation, the LBW rate is convenient and actually to some extent characterizes both mean and SD. Therefore we choose to use the rate of term LBW (i.e., gestational period ≥37 weeks AND birth weight < 2500g) as another measurement of adverse reproductive outcome in this study.

Birth Data and Rate Calculation—We obtained birth data from New Hampshire birth certificates for 1997-2009 (N=187,851), provided by New Hampshire Department of Health

and Human Services (NH DHHS). Each record in the dataset is for an infant and contains information about: 1) the infant, including birth date, gestational age, sex, birth weight, plurality, and birth order; and 2) the mother, including age, residential town, and zip code at delivery.

Prior to the analysis, we removed those records of mothers who were not residents of New Hampshire towns, which account for about 1% of the original records. We then removed those records with a plurality value > 1 (i.e., twins and triplets), which account for about 3% of all records. We also removed those records with apparent invalid or missing values on gestational period, maternal age, and birth weight, which account for less than 1% of all records. After these processes, a total of 177,995 records remained and were used in the following analyses.

Rate of Preterm Birth: Among the 177,995 usable records, 12,501 have a gestational period < 37 weeks and were identified as preterm births. We stratified the data into detailed categories of maternal age and calculated the preterm birth rate for each category (the upper part of Table 1). The calculation reveals a step at maternal age = 20. We hence grouped the detailed categories into two larger categories: maternal age < 20 and maternal age ≥ 20 (the lower part of Table 1). Although the categories of maternal age ≥ 40 have greater rates, their relatively small counts for both preterm births and all births may lead to statistical instability, and therefore were grouped into the category of maternal age ≥ 20. We performed the following analyses separately for the two strata to address the influence of maternal age.

Rate of term-LBW: From the 177,995 usable records, we identified 2,651 LBW cases with full term pregnancies (i.e., gestational period ≥ 37 weeks AND birth weight < 2500g). Similar to the process with the preterm birth data, we stratified the data into detailed categories of maternal age and calculated the term LBW rate for each (the upper part of Table 2). The calculation reveals a clear step at maternal age = 24. We hence grouped the detailed categories into two larger categories: maternal age < 25 and maternal age ≥ 25, and performed the following analyses separately for these two strata. Similar to the preterm birth data, we grouped the categories of maternal age ≥ 40 into the category of maternal age ≥ 25, due to their relatively small counts.

Population and Socioeconomic Data—The population data used in the rate calculations described above is from the US Census 2010 data (<http://www.census.gov/>). In this study, however, we did not include the factor of race/ethnicity, because the birth certificates do not contain such information. According to the Census 2010 data, non-white females in NH only account for 7% of the female population within the age range 15–49. Thus, we assumed that any race/ethnicity effect is negligible. The risk of birth problems may also be affected by socioeconomic status. We collected town-level income data from the New Hampshire Office of Energy and Planning (www.nh.gov/oep).

Arsenic Data—The exposure data used in this study are the modeled probabilities of arsenic occurrence (at thresholds of 1, 5, and 10 µg/L) in private wells that tap groundwater from bedrock aquifers in New Hampshire. This probability of finding arsenic at a location

above a given threshold was estimated using multivariate logistic regression models (“probability models”) developed for New Hampshire [28]. The probability models were developed using measurements of arsenic from public and private wells as the dependent (or predicted) variable, and using a variety of geologic, geochemical, hydrologic, and anthropogenic data as the independent variable (predictor) [28-31].

Probability models for predicting arsenic concentrations that were greater than or equal to 1, 5, and 10 µg/L in bedrock wells were developed in order to produce individual threshold-level probability maps. These three thresholds were chosen because they represent common arsenic reporting levels in groundwater in the State and because they are considered to be possibly relevant concentrations for exposure estimation in terms of potential human-health outcomes. Also, the current USEPA maximum contaminant level (MCL), the standard for safe drinking water with which public water supplies in the United States must comply, is 10 µg/L. The multivariate logistic regression techniques used to generate the probability estimates are well suited for modeling censored dependent-variable data—data reported as “less than” some laboratory reporting limit—because data that are below reporting limits can be used directly without having to modify or substitute values [32-34]. The well-water arsenic concentration data (dependent data) include censored data that were reported as below laboratory reporting levels (LRLs). The model takes the form:

$$P [y=1|x] = \frac{e^{(\beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_k x_k)}}{1 + e^{(\beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_k x_k)}} \quad (1)$$

where P is the probability of observing the event, y is an indicator (threshold) variable (“y=1” denoting an event or measurement greater than or equal to a specific value (such as 1, 5, and 10 µg/L), and “y=0” denoting a non-event or measurement less than a specific threshold), where x_1, x_2, \dots, x_k are explanatory or independent variables, and where $\beta_0, \beta_1, \dots, \beta_k$ are unknown parameters (coefficients) to be estimated. The exponential of a parameter, $\exp(\beta_i)$, specifies the proportional increase in the odds of an arsenic concentration being above the modeled threshold per unit increase in the explanatory variable. Threshold values of 1, 5, and 10 µg/L were modeled to identify areas of the State where the probabilities are high for finding low-level (greater than or equal to 1 µg/L) and high-level (greater than or equal to 10 µg/L) arsenic contamination in groundwater. Standard model testing and performance metrics were evaluated and are described in detail elsewhere [28].

We tested data representing concentrations of arsenic in 1,715 wells (dependent variable) to develop the models, along with more than 250 independent variables, all developed in a geographic information system (GIS), and representing geologic, hydrologic, demographic, and land-use and land-cover features. The final models were dominated by geologic and geochemistry variables but also included variables such as population density, precipitation, groundwater recharge, land use, and proximity to waste sites [28].

The probability of having arsenic concentrations exceeding 1, 5, and 10 µg/L in groundwater was variable across the state. Generally, high probabilities of arsenic greater than 5 or 10 µg/L were limited to southeastern New Hampshire. However, high probabilities that groundwater from bedrock aquifers would exceed 1 µg/L were widespread across New

Hampshire. In fact, nearly half of the State was classified as having at least a 50% chance of having arsenic greater than or equal to 1 $\mu\text{g/L}$. High probabilities of arsenic greater than or equal to 5 and 10 $\mu\text{g/L}$ were predicted primarily in the southeastern counties of Merrimack, Strafford, Hillsborough, and Rockingham—the counties that are home to about 75% of the State's population.

The original USGS arsenic data are in the format of GIS raster layers, with cell size = 30 m. To match the LBW data at the town level, the cell-level data were aggregated to town level using the Zonal Statistics tool of ArcGIS, i.e., the average of the values of all the cells falling into a town is used as the representative value of that town (Figure 1).

Methods

Rate Smoothing

To statistically stabilize the rates, we applied a locally-weighted-average smoothing to the original rates. Locally-weighted-average methods smooth the rate of an areal unit (in our case, a town) by averaging all the rates of the units in its neighborhood, during which each rate is weighted by its associated background value (in our case, number of births) [35,36]. The specific method we implemented was proposed and justified by Shi et al. [37], and is different from conventional locally-weighted averaging methods. This method 1) employs a user-specified *background* value rather than a constant geographic distance to define the neighborhood for smoothing, which makes the statistical stability explicit and controllable; 2) generates the neighborhood by creating a buffer around the polygon, rather than about the centroid of the polygon, which takes into account the size and shape of the polygon; and 3) if the neighborhood encloses only part of a polygon, the weight of that polygon will be proportionally determined, allowing a more accurate estimation of contribution of each polygon than an *in-or-out* strategy.

To address the subjectivity in determining the threshold for defining the neighborhood, Shi et al. proposed a strategy that calculates a series of smoothing results using different thresholds. For each of these results, the overall variance of the smoothed rates is calculated and plotted [37]. It is expected that the variance values become stable as the threshold increases, and the turning point on the plot where the variance value starts to level out is considered an indication of the optimal threshold. We used this strategy in the current study to identify optimal thresholds.

Correlation Calculation

To detect the spatial association between birth problems and groundwater arsenic, we calculated a Pearson's correlation coefficient (r) between the town-level preterm birth rates and the groundwater arsenic levels, and between the term-LBW rates and the groundwater arsenic levels. For both preterm births and term-LBW, the calculation was performed on the original rates and a series of smoothed rates with different thresholds. The calculation was performed for the different maternal age strata separately.

For each stratum, we started the smoothing with a threshold = 100 births, i.e., the buffer around each town polygon was expanded until it enclosed 100 births. With this threshold,

those towns with more than 100 births during the study period would not be smoothed. We kept increasing the threshold, using 100 as the increment, to get smoother results. Figure 2 shows the maps of the original rates and the smoothed rates with a threshold = 2,200 side by side for a visual comparison. These maps are for younger mothers in the term-LBW analysis.

With the map of smoothed rates from a threshold, we calculated r between the rates and arsenic level, as well as the variance of rates across the entire state. As an example, Figure 3 shows the results of these calculations for younger mothers in the term-LBW analysis. Figure 2 indicates that for younger mothers, the correlation between the term-LBW rate and groundwater arsenic almost monotonically increases after the only major drop associated with the starting threshold (100 births). With a threshold of 3,500 births, r reaches 0.3. We stopped at 3,500 to avoid oversmoothing. In fact, the variance starts to stabilize when the threshold = 2,000 and becomes very small after 2,500.

To eliminate possible impacts of population and income on the risk of birth problems, following Ayotte et al. [29], we also applied linear regression using the disease rate as the dependent variable, and population or income as the independent variable. If the rate was related to population or income, we calculated residuals of the dependent variable, and then calculated correlations between the residuals and the arsenic values.

To explore the local variation of the spatial association, we also calculated the correlation coefficient for each of the 10 counties of New Hampshire.

RESULTS

Preterm Birth

For the stratum of maternal age < 20 (Table 3), statewide the unsmoothed town-level preterm birth rates have slightly negative r values for all three arsenic data layers, as well as for the town household median income value. However, the smoothing dramatically changes this situation. Even with a relatively small smoothing threshold of 500 births, r values all become positive with considerable magnitudes. Generally, the higher the degree of smoothing, the higher the r value. The largest r value, 0.70, occurs between the town mean arsenic level based on the data of probability > 10 $\mu\text{g/L}$ and the preterm birth rate smoothed with a threshold = 2000 births (the largest threshold used in this analysis). Among the three arsenic data layers, the ones of probability > 5 $\mu\text{g/L}$ and > 10 $\mu\text{g/L}$ have stronger positive associations with the preterm birth rate than the one of probability > 1 $\mu\text{g/L}$. Among the town minimum, maximum, and mean for the arsenic probability, generally the mean has the highest r value, whereas the minimum has the lowest. Unexpectedly, the smoothed rates have fairly considerable positive correlations with the town median household income, i.e., higher rates tend to be associated with higher income values. The adjustment by income consistently lowers down the r values, indicating that the income and arsenic may have an association to a certain extent. However, even after the adjustment, the positive association between preterm birth and groundwater arsenic for this group of mothers in New Hampshire is still considerable.

At the county level, generally the r values progressively become more positive as the arsenic measurement becomes stricter, from probability $> 1 \mu\text{g/L}$, $> 5 \mu\text{g/L}$, to $> 10 \mu\text{g/L}$ (Figure 4). For the data of probability $> 10 \mu\text{g/L}$, a majority of r values are positive. This progressive variation is most dramatic for the five most populous counties (in terms of population density) in New Hampshire, including Hillsborough, Rockingham, Stafford, Merrimack, and Belknap (the five left-most counties in Figure 4), which account for 78% of the state's total population, and geographically cluster at the southwest corner of the state that is close to the greater Boston area. For the data of probability $> 1 \mu\text{g/L}$, four of these five counties have dominantly negative r values, but for the data of probability $> 10 \mu\text{g/L}$, most r values of these counties are positive. Another noteworthy finding is that in most cases, the smoothing “helps” increase positiveness.

For the stratum of maternal age ≥ 20 , the r values are much smaller, compared with their counterparts of the younger mothers (Table 4). While it is hard to claim any significant association based on these r values, it seems, however, that the general pattern of them is similar to that of the younger mothers. The data of probability $> 5 \mu\text{g/L}$ and probability $> 10 \mu\text{g/L}$ have stronger positive associations with the preterm birth rate than the data of probability $> 1 \mu\text{g/L}$. In fact, all r values based on the $5 \mu\text{g/L}$ and $10 \mu\text{g/L}$ data are consistently positive, although small. Again, the smoothing generally helps increase positiveness for r . The largest r value, 0.19, occurs between the town-level mean for probability $> 10 \mu\text{g/L}$ and the preterm birth smoothed with the largest threshold (3,500 births). However, in this stratum the preterm rate does not appear to have considerable associations with household median income, therefore we did not calculate the income-adjusted r .

At the county level, the progressively change of r values along the three arsenic data layers is still obvious (Figure 5). For the data of probability $> 10 \mu\text{g/L}$, a majority of r values are positive. This time, the role of smoothing is controversial. For example, for counties of Cheshire and Carroll, smoothing makes r values stably and increasingly positive, while for counties of Rockingham, Strafford, and Belknap, smoothing reduces the positiveness.

Term Low Birth Weight

The results of term-LBW are generally weaker and less consistent than those of the preterm birth. For the stratum of maternal age < 25 years, statewide positive r values are dominant across all tests performed (Table 5), but some patterns are different from those of the preterm birth. First, the rank of the three arsenic data layers is reversed and this time the data of probability $> 1 \mu\text{g/L}$ have stronger association with the rates than the other two. Second, among the town minimum, maximum, and mean, the minimum consistently has the highest positive r values than the other two. The smoothing still helps increase the positiveness. The r values between the rates and income are fairly small, and therefore we did not calculate the income-adjusted r .

Figure 6 shows the county-specific r values between the town minimum arsenic probability value and the term-LBW rate. While statewide more positive r is associated with the arsenic data of $1 \mu\text{g/L}$, at the county level the dominance of positive r values is more obvious with the data of $10 \mu\text{g/L}$. It seems that the inverted results of the Merrimack County with the data

of 1 $\mu\text{g/L}$ and 10 $\mu\text{g/L}$ have caused this controversy. The geographic pattern largely maintains: The three most populous counties (in terms of population density), including Hillsborough, Rockingham, and Strafford, accounting for 62% of the states' population and clustering near the great Boston area, generally have positive r values across the three arsenic data layers. The smoothing, again, in most cases helps increase the positiveness.

For the stratum of maternal age ≥ 25 years, statewide r is dominated by negative values, a few are fairly considerable (e.g., $r = -0.36$ for the town maximum of probability $> 10 \mu\text{g/L}$ and the rates smoothed with a threshold = 2,500 births), although most are very small (Table 6). Along the line from 1 $\mu\text{g/L}$, 5 $\mu\text{g/L}$, to 10 $\mu\text{g/L}$, the negativeness generally increases, especially for the town maximum and mean. The effect of smoothing does not have an obvious pattern. The rates, however, have non-negligible negative associations with town median household income (higher rates tend to be associated with lower incomes). The adjustment by the income unexceptionally reduces the negativeness of the r values, indicating that income may be a confounder that is overshadowing the effect of groundwater arsenic.

The county-specific correlations for the stratum of maternal age ≥ 25 years are weak and inconsistent overall. Figure 7 shows the county-specific r values between the income-adjusted term LBW rates and the town minimum arsenic probability. The r values are generally small (no matter positive or negative), compared with the stratum of younger mothers. Negative r values prevail, in terms of the number of counties, with the data of 1 $\mu\text{g/L}$ and 10 $\mu\text{g/L}$; with the data of 5 $\mu\text{g/L}$, it is a mix. The most populous county, Hillsborough, maintains to be generally positive across the three arsenic data layers.

Discussion

We found evident spatial associations between two adverse human reproductive outcomes, including preterm birth and term LBW, and groundwater arsenic in New Hampshire, US. However, the properties of these associations vary, in terms of degree of association and sensitivity to different representations of arsenic level. Generally, preterm birth has a stronger spatial association with groundwater arsenic than term LBW, suggesting an inconsistency in the impact of arsenic on the two reproductive outcomes, and confirming the necessity to distinguish preterm births and term births in this kind of analysis. For both reproductive outcomes, younger maternal age has stronger spatial associations with groundwater arsenic. Especially for term LBW, while a positive spatial association between LBW and arsenic level is observed for maternal < 25 , the association is unclear for maternal age ≥ 25 . However, an initial exploration with town median household income suggests that for the stratum of maternal age ≥ 25 the effect of groundwater arsenic might have been shadowed by socioeconomic or other factors.

In this study, we treated town median household income as a confounding factor. However, the associations between the reproductive outcome and household income are fairly variable across outcomes and maternal ages. The preterm birth with maternal age < 20 has a stronger and unexpected positive association with the household income, and the adjustment by household income consistently reduces the positiveness in the association between the

outcome and the arsenic, suggesting that the spatial distribution of household income may co-vary with groundwater arsenic to some extent in NH. The preterm birth with maternal age ≥ 20 has a slight but negative association with the household income, as well as a much weaker positive association with arsenic, which can be interpreted as that the negative effect of income and the positive effect of arsenic have cancelled each other to some extent. For term LBW, the correlation between the outcome and household income is noticeable for maternal age ≥ 25 . The adjustment by income unexceptionally reduces the negativeness in the r values between LBW rates and arsenic levels. For the stratum of maternal age < 25 , the correlation between the LBW rate and the income is very weak. While these findings are seemingly variable, they can have a fairly consistent interpretation: younger mothers, who may have not established a career and/or a stable income, may be less sensitive to the expected negative effect of an economical variable, especially measured at a highly aggregated level, and as a result the impact of environmental hazards on them might be easier to detect. For older mothers, income might be a stronger independent variable in the equation. This may also indicate that town-level income data may not well represent the economic status of younger mothers.

This study reveals that a smoothing process may have a considerable effect on detection of spatial association. In the analysis of preterm birth and in the analysis of younger mothers with term LBW, it appears that more statistically stable rates (i.e., more smoothing) help reveal potential associations between the reproductive outcomes and the groundwater arsenic. We are aware that for less populous areas such as northern New Hampshire, there is a greater risk of over-smoothing, i.e., the smoothed rates may not correctly reflect the local variability of the disease risk, and in turn, may affect the reliability of the detected association between the disease and environmental exposures.

We find that the spatial associations between the reproductive outcomes and the groundwater arsenic have spatial variation in New Hampshire, and this regional pattern is consistent, as expected, for the two reproductive outcomes we examined. Among the 10 counties of NH, the several most populous counties usually have stronger spatial associations between birth problems and groundwater arsenic than the others. Geographically, these counties are clustered in the southeastern corner of New Hampshire and are more proximate to the greater metropolitan Boston area. There are at least three factors that may have contributed to the stronger correlations they possess: 1) these counties are the most populated, having a majority of the state's population, and, as such, provided a larger sample size with more stable disease rates than in other parts of the state, which may have better reflected the actual influence of groundwater arsenic on reproductive outcomes; 2) some parts of these counties have the highest arsenic values in the USGS modeled data, and the high arsenic exposure levels may have increased the detectability of an association between the reproductive outcomes and groundwater arsenic; and 3) compared with the rest of the state, the arsenic values vary the most in this region, which also may have facilitated the detection of its spatial association with LBW.

The New Hampshire population uses either public or private water supplies. There is no arsenic regulatory requirement for private water supplies, which results in some having fairly high arsenic concentrations. In contrast, public supplies, by law, must have arsenic

monitored and controlled. A limitation of this study is that we did not have precise information to distinguish different water sources in different places. Partially because of this, similar to most geospatial analyses in health studies, the goal of this study has been set to be “exploration”. We were using geospatial analysis and the best available data to explore if there is a possibility that arsenic in groundwater has an association with adverse reproductive outcomes. To have accurate and precise information about people's source of drinking water requires much more extensive and expensive investigations, which is outside the scope of the study presented by this paper. It should also be noted that, as a fairly rural state, New Hampshire's 40% population using private wells disproportionately occupy a much larger geographic area than the other 60% of the population. Nevertheless, it will be of interest to make the distinction between private and public water supplies in future data collection and analyses.

In this study we used three GIS data layers of modeled probability of arsenic exceeding certain concentration levels, including 1 $\mu\text{g/L}$, 5 $\mu\text{g/L}$, and 10 $\mu\text{g/L}$. While the results based on the three data layers are generally consistent, the distinction is noticeable. For the preterm birth analysis, the one with the highest bar, probability > 10 $\mu\text{g/L}$, tend to bring about strongest positiveness. For the term LBW analysis, the one with the lowest bar, probability > 1 $\mu\text{g/L}$, sometimes is more “positive” than the others. What is also noteworthy is the different “performances” of the three representations of town-level arsenic: minimum, maximum, and mean. For the preterm birth analysis, the town mean provides the most positive r values, and for the term LBW analysis, the town minimum slightly outperforms the other two. A general lesson learned from these findings is that in geospatial analysis for environmental health studies, a thorough consideration and exploration of different data, representations, and parameter settings is necessary.

In principle, the analysis should take into account all confounding factors, subject to availability of data. Maternal age is the only confounding factor we have data at the individual level. Race/ethnicity, a known confounding factor, was not taken into account of this study, due to lack of data of this variable at the individual level. Fortunately, the New Hampshire population is over 95% Caucasian, making us to assume that the impact of this disadvantage is minimal. We analyzed the town-level income data as our best-possible effort so far to address socioeconomic factors. Nutritional factors influence biomarker concentrations of arsenic, and could potentially be confounders. This is a limitation of our analysis, but we are limited by data in this study, and will certainly take them into account whenever data allow.

Whereas the correlations reported in this paper appear to indicate a relation between the two reproductive outcomes and groundwater arsenic, it does not lead to the conclusion that arsenic from bedrock wells is the true cause of the relation. It is possible that other correlates or combinations of factors that follow a similar pattern to that of arsenic is responsible for the relation we observed. Future investigations of these and other relations may provide additional insight into understanding the effect of exposure to arsenic from private wells in the region.

Finally, we consider it is worth emphasizing the fact that in this study we found a relation to low-to-moderate level arsenic. The finding is novel also because very little is known about adverse reproductive outcomes and arsenic exposure in a US population. It has international value because there are so many parts of the world with low-to-moderate levels that are not on the general arsenic radar screen.

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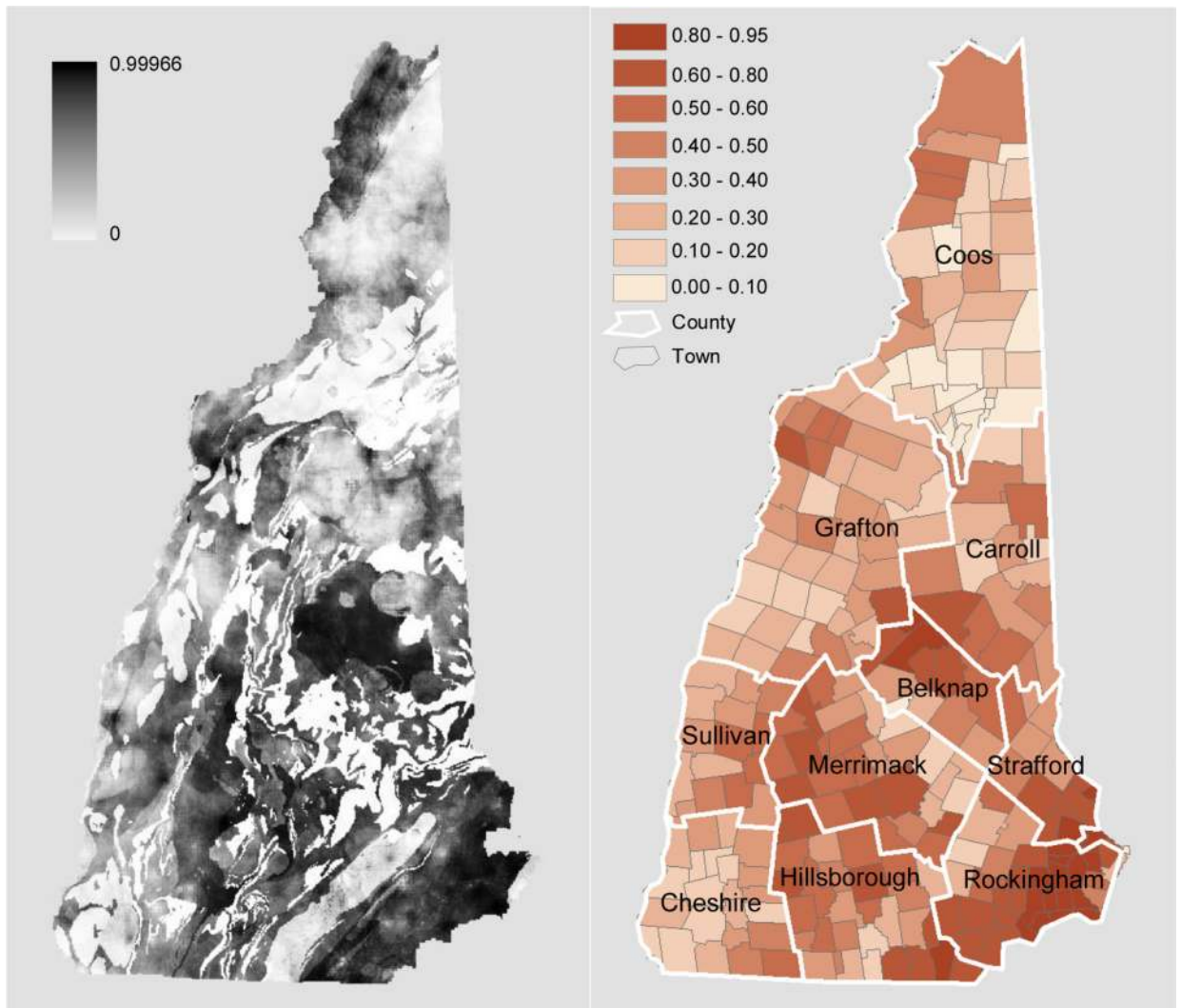


Figure 1.

The USGS modeled groundwater arsenic value in New Hampshire, represented as the probability of arsenic concentration exceeding a certain level. The left map is the original USGS raster data showing the probability of arsenic concentration > 1 ug/L; the right map is the town level mean values calculated from the left map using the *zonal statistics* tool of ArcGIS.

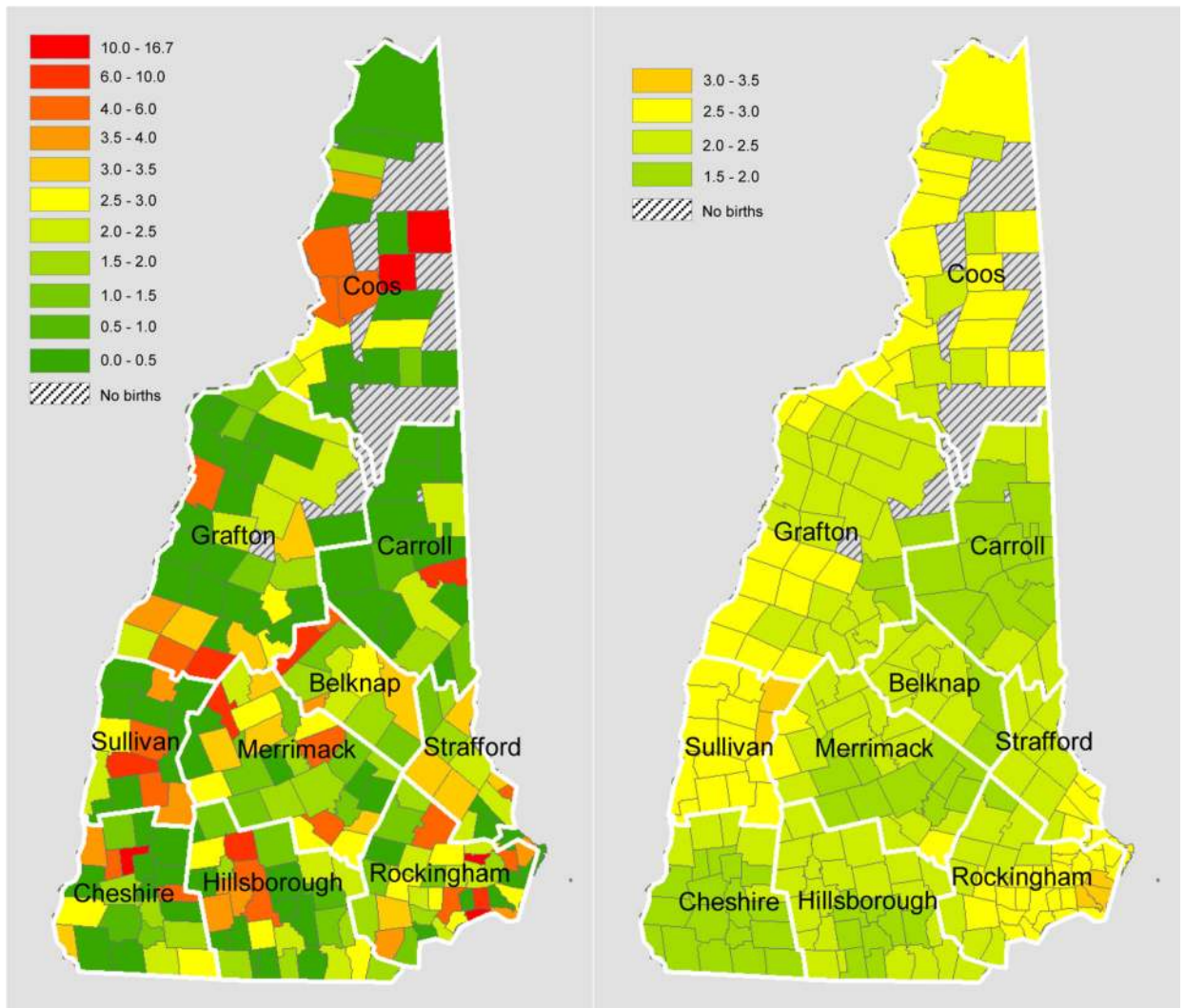


Figure 2. New Hampshire town-level LBW rates for maternal age < 25: The left map displays the original rates; the right map displays the smoothed rates with smoothing threshold = 2,200 births.

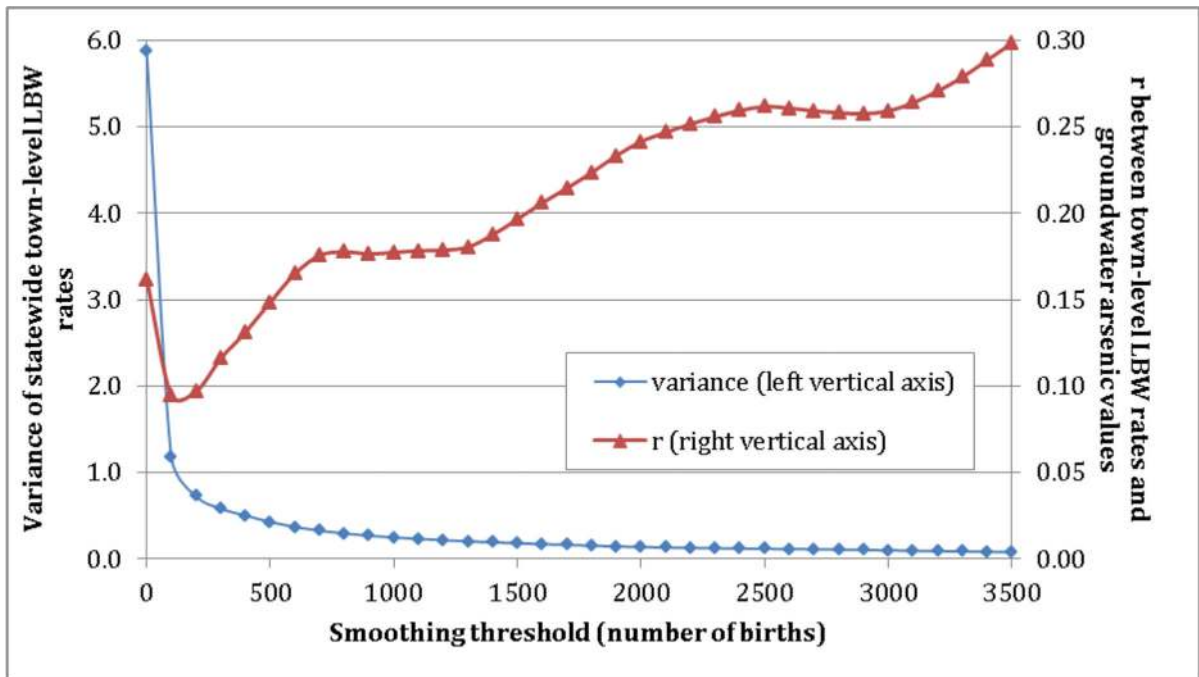


Figure 3. Variance of New Hampshire town-level LBW rates for maternal age < 25 and correlation between the rates and groundwater arsenic against smoothing threshold.

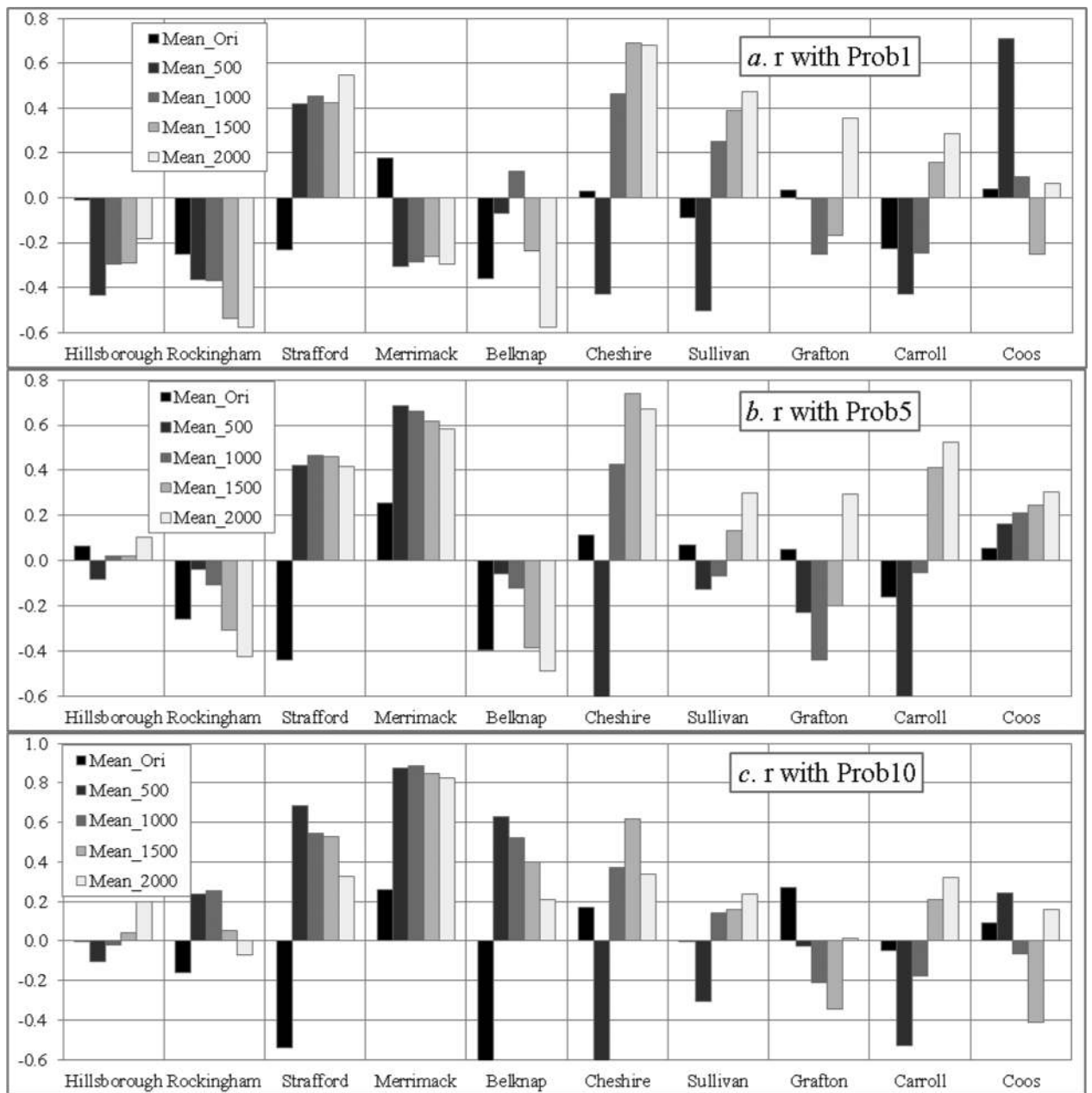


Figure 4.

County-specific correlation coefficient (r) between town-level preterm birth rate for maternal age < 20 and groundwater arsenic in New Hampshire. Notes: *Prob1*, *Prob5*, and *Prob10* denote three GIS data layers of modeled groundwater arsenic level, representing probability of arsenic > $1\mu\text{g/L}$, $5\mu\text{g/L}$, and $10\mu\text{g/L}$, respectively; *Mean_Ori* denotes the r between the town-level mean probability of arsenic level and the original preterm birth rate; *Mean_500* denotes the r between the town-level mean probability of arsenic level and the preterm birth rate smoothed with threshold = 500 births; and so on.

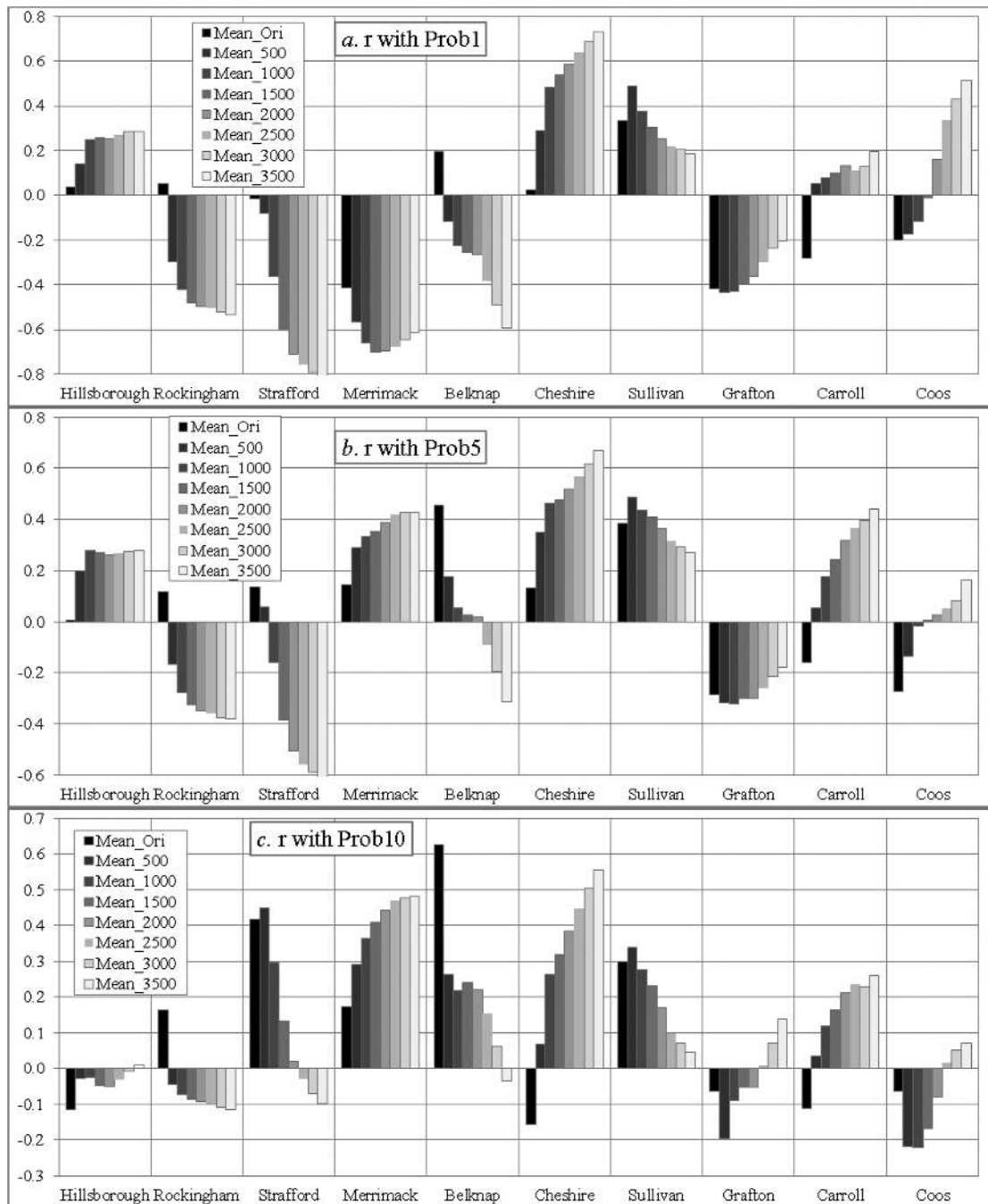


Figure 5.

County-specific correlation coefficient (r) between town-level preterm birth rate for maternal age ≥ 20 and groundwater arsenic in New Hampshire. Notes: *Prob1*, *Prob5*, and *Prob10* denote three GIS data layers of modeled groundwater arsenic level, representing probability of arsenic $> 1\mu\text{g/L}$, $5\mu\text{g/L}$, and $10\mu\text{g/L}$, respectively; *Mean_Ori* denotes the r between the town-level mean probability of arsenic level and the original preterm birth rate; *Mean_500* denotes the r between the town-level mean probability of arsenic level and the preterm birth rate smoothed with threshold = 500 births; and so on.

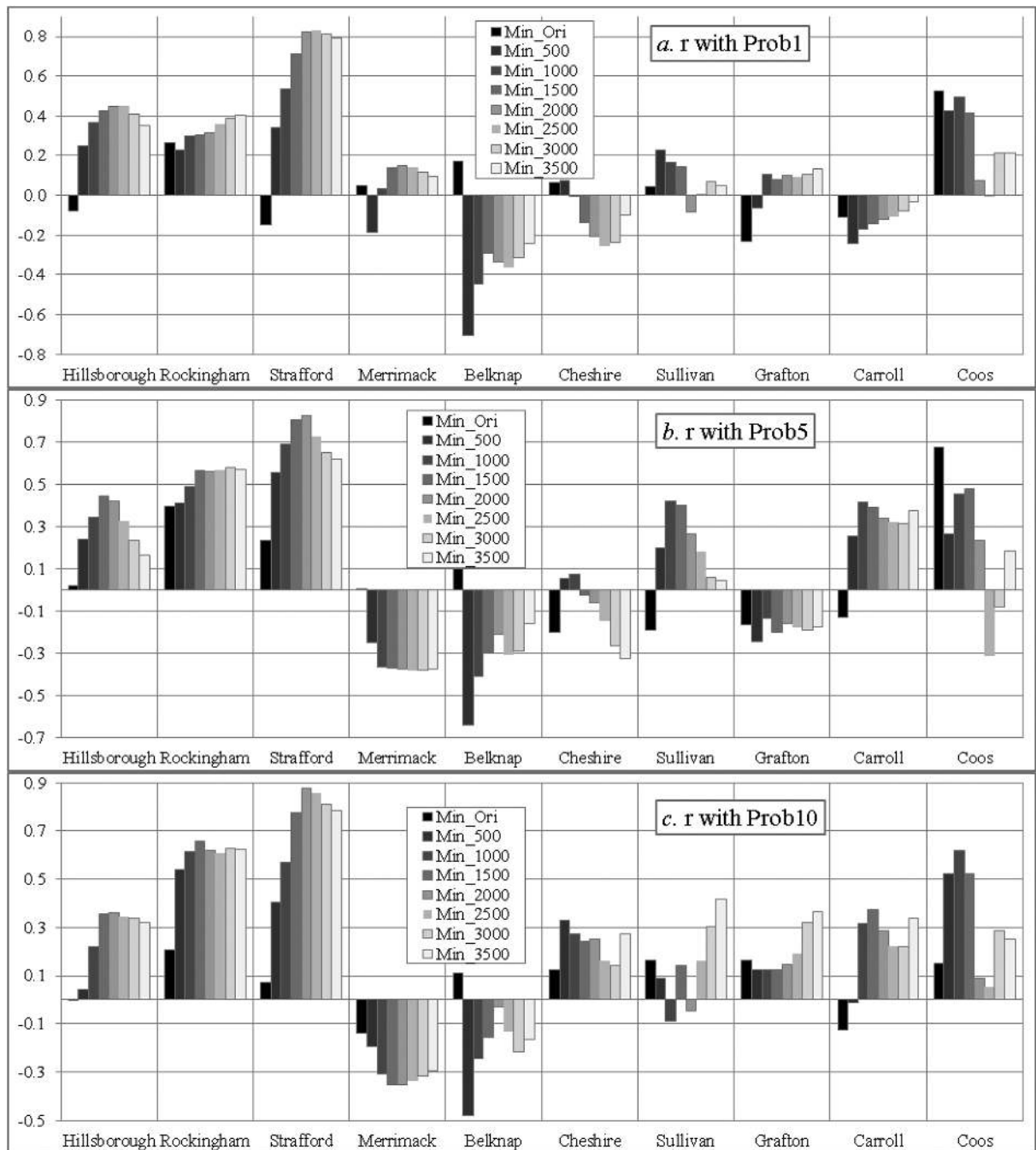


Figure 6.

County-specific correlation coefficient (r) between town-level term-LBW rate for maternal age < 25 and groundwate arsenic in New Hampshire. Notes: *Prob1*, *Prob5*, and *Prob10* denote three GIS data layers of modeled groundwate arsenic level, representing probability of arsenic > 1 μ g/L, 5 μ g/L, and 10 μ g/L, respectively; *Min_Ori* denotes the r between the town-level minimum probability of arsenic level and the original term-LBW rate; *Minimum_500* denotes the r between the town-level mean probability of arsenic level and the term-LBW rate smoothed with threshold = 500 births; and so on.

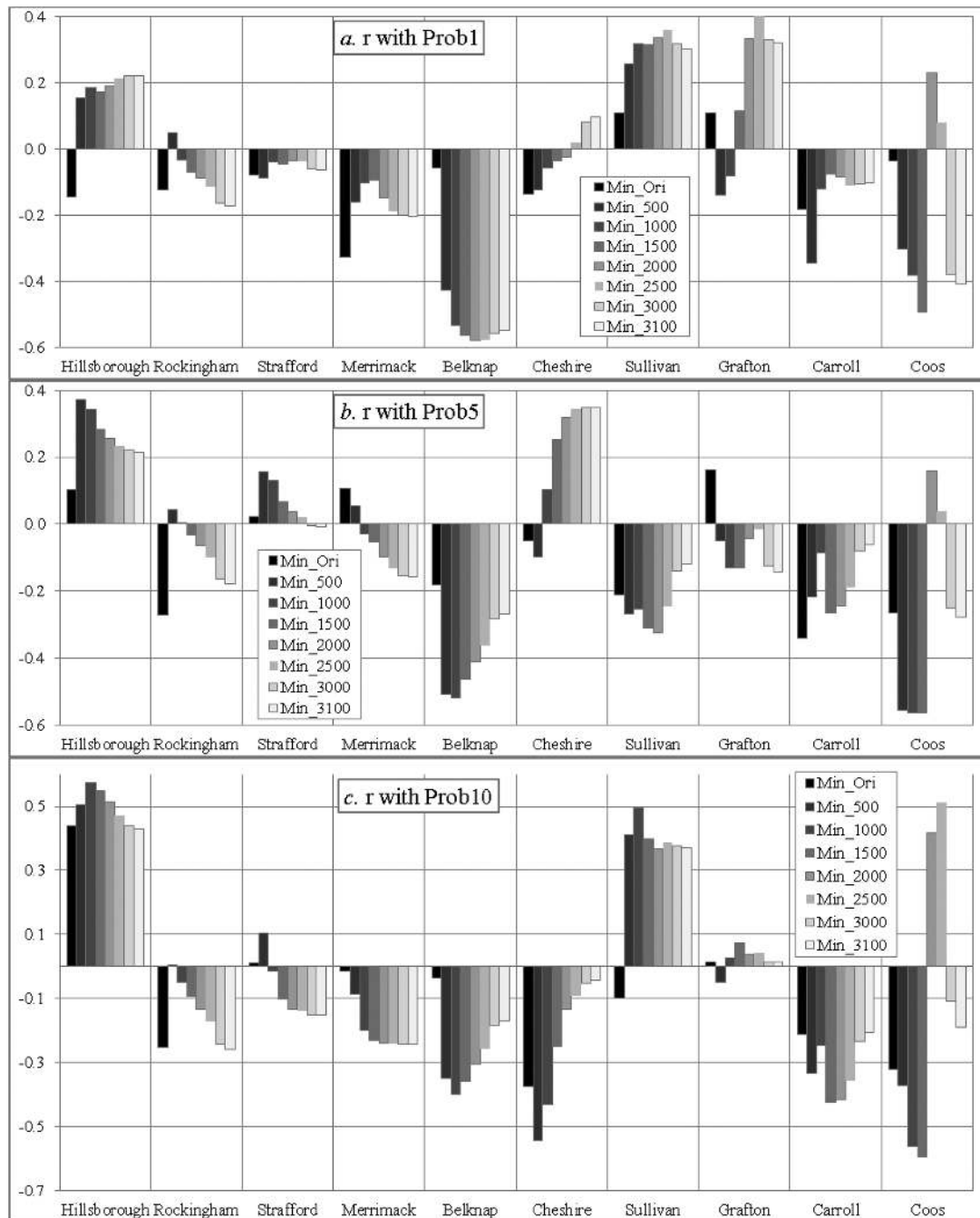


Figure 7.

County-specific correlation coefficient (r) between town-level term-LBW rate for maternal age ≥ 25 and groundwater arsenic in New Hampshire. Notes: *Prob1*, *Prob5*, and *Prob10* denote three GIS data layers of modeled groundwater arsenic level, representing probability of arsenic $> 1\mu\text{g/L}$, $5\mu\text{g/L}$, and $10\mu\text{g/L}$, respectively; The term-LBW rates used in this figure have been adjusted by the town median household income of 1999; *Min_Ori* denotes the r between the town-level minimum probability of arsenic level and the income-adjusted original term-LBW rate; *Min_500* denotes the r between the town-level minimum

probability of arsenic level and the income-adjusted term-LBW rate smoothed with threshold = 500 births; and so on.

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Table 1

Preterm birth ratio by maternal age category, New Hampshire, 1997-2009.

38. Maternal Age	Preterm Births	All Births	Ratio
< 20	11792	1034	0.0877
20-24	34272	2487	0.0726
25-29	49665	3317	0.0668
30-34	51691	3400	0.0658
35-39	25368	1783	0.0703
40-44	4975	456	0.0917
≥45	232	24	0.1034
Two-Category Stratification:			
<20	11792	1034	0.0877
>=20	166203	11467	0.0690

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Table 2

Low Birth Rate (LBW) ratio by maternal age category, New Hampshire, 1997-2009.

Age	LBW Infants	All Births	Ratio
<20	274	10569	0.0259
20-24	685	31207	0.0220
25-29	665	47762	0.0139
30-34	615	47225	0.0130
35-39	325	22990	0.0141
40-44	80	4404	0.0182
>=45	7	198	0.0354
Two-Category Stratification:			
<25	959	41776	0.0230
>=25	1692	122579	0.0138

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Table 3

Correlation Coefficient (r) between town-level preterm birth rate for maternal age < 20 and groundwater arsenic in New Hampshire.

		Original_Rate	Smoothed_500	Smoothed_1000	Smoothed_1500	Smoothed_2000
<i>Prob1</i>	<i>Town_Min</i>	-0.09	0.25 (0.14)	0.24 (0.14)	0.28 (0.17)	0.31 (0.15)
	<i>Town_Max</i>	-0.01	0.37 (0.30)	0.29 (0.22)	0.36 (0.29)	0.43 (0.32)
	<i>Town_Mean</i>	-0.08	0.26 (0.15)	0.23 (0.13)	0.30 (0.20)	0.39 (0.23)
<i>Prob5</i>	<i>Town_Min</i>	-0.06	0.47 (0.37)	0.44 (0.35)	0.50 (0.42)	0.56 (0.42)
	<i>Town_Max</i>	-0.13	0.39 (0.26)	0.48 (0.37)	0.57 (0.46)	0.64 (0.47)
	<i>Town_Mean</i>	-0.07	0.46 (0.33)	0.48 (0.36)	0.57 (0.45)	0.65 (0.47)
<i>Prob10</i>	<i>Town_Min</i>	-0.06	0.47 (0.40)	0.46 (0.38)	0.47 (0.40)	0.49 (0.38)
	<i>Town_Max</i>	-0.07	0.43 (0.31)	0.41 (0.30)	0.51 (0.41)	0.61 (0.45)
	<i>Town_Mean</i>	-0.04	0.57 (0.46)	0.56 (0.47)	0.63 (0.54)	0.70 (0.55)
<i>Median household Income</i>		-0.06	0.34	0.30	0.39	0.47

Notes: *Prob1*, *Prob5*, and *Prob10* denote three GIS data layers of modeled groundwater arsenic level, representing probability of arsenic > 1µg/L, 5µg/L, and 10µg/L, respectively; *Town_Min*, *Town_Max*, and *Town_Mean* denote town-level minimum, maximum, and mean for the modeled arsenic probability values, respectively; *Smoothed_500* etc. denote the smoothed rates; e.g., *smoothed_500* denotes the town-level preterm birth rate smoothed from the *Original Rate* using a threshold that the neighborhood of smoothing must enclose at least 500 births; The value inside the parentheses is r adjusted by town-level income (i.e., r between the arsenic value and the residual to the preterm birth-income linear regression); The bottom line contains the r values between the preterm birth rate and the town median household income in 1999; Since the original rate has very weak correlation with the income value, the income-adjusted r for the original rate was not calculated.

Table 4

Correlation Coefficient (r) between town-level preterm birth rate for maternal age ≥ 20 and groundwater arsenic in New Hampshire.

	Original Rate	Smoothed 500	Smoothed 1000	Smoothed 1500	Smoothed 2000	Smoothed 2500	Smoothed 3000	Smoothed 3500
<i>Prob1</i>	<i>Town_Min</i>	0.00	-0.04	-0.06	-0.06	-0.05	-0.05	-0.05
	<i>Town_Max</i>	-0.05	-0.02	0.01	0.04	0.07	0.11	0.13
	<i>Town_Mean</i>	-0.10	-0.10	-0.09	-0.06	-0.02	0.01	0.03
<i>Prob5</i>	<i>Town_Min</i>	0.04	0.10	0.09	0.10	0.11	0.12	0.14
	<i>Town_Max</i>	0.01	0.05	0.08	0.10	0.13	0.14	0.15
	<i>Town_Mean</i>	0.01	0.06	0.08	0.10	0.13	0.14	0.18
<i>Prob10</i>	<i>Town_Min</i>	0.04	0.06	0.06	0.06	0.07	0.07	0.07
	<i>Town_Max</i>	0.03	0.00	0.01	0.02	0.04	0.05	0.09
	<i>Town_Mean</i>	0.05	0.08	0.10	0.11	0.13	0.15	0.19
<i>Median household Income</i>	-0.12	-0.15	-0.13	-0.11	-0.08	-0.06	-0.04	-0.03

Notes: *Prob1*, *Prob5*, and *Prob10* denote three GIS data layers of modeled groundwater arsenic level, representing probability of arsenic $\geq 1\mu\text{g/L}$, $5\mu\text{g/L}$, and $10\mu\text{g/L}$, respectively; *Town_Min*, *Town_Max*, and *Town_Mean* denote town-level minimum, maximum, and mean for the modeled arsenic probability values, respectively; *Smoothed_500* etc. denote the smoothed rates; e.g., *smoothed_500* denotes the town-level preterm birth rate smoothed from the *Original Rate* using a threshold that the neighborhood of smoothing must enclose at least 500 births; The bottom line contains the r values between the preterm birth rate and the town median household income in 1999; Since the preterm birth rate has very weak correlation with the income value, we did not calculate the income-adjusted r values.

Table 5

Correlation Coefficient (r) between town-level term-LBW rate for maternal age < 25 and groundwater arsenic in New Hampshire.

	Original Rate	Smoothed 500	Smoothed 1000	Smoothed 1500	Smoothed 2000	Smoothed 2500	Smoothed 3000	Smoothed 3500
<i>Prob1</i>	<i>Town_Min</i>	0.19	0.25	0.30	0.35	0.37	0.40	0.44
	<i>Town_Max</i>	0.00	-0.02	-0.01	0.00	0.01	0.01	0.03
	<i>Town_Mean</i>	0.16	0.19	0.21	0.25	0.26	0.26	0.30
<i>Prob5</i>	<i>Town_Min</i>	0.15	0.18	0.23	0.27	0.28	0.29	0.34
	<i>Town_Max</i>	0.02	0.02	0.04	0.08	0.08	0.09	0.14
	<i>Town_Mean</i>	0.10	0.12	0.16	0.21	0.21	0.22	0.27
<i>Prob10</i>	<i>Town_Min</i>	0.23	0.25	0.30	0.34	0.36	0.38	0.43
	<i>Town_Max</i>	-0.12	-0.18	-0.18	-0.16	-0.16	-0.15	-0.09
	<i>Town_Mean</i>	0.02	-0.02	0.00	0.04	0.04	0.05	0.11
<i>Median household Income</i>	-0.02	-0.03	0.02	0.05	0.10	0.12	0.12	0.18

Notes: *Prob1*, *Prob5*, and *Prob10* denote three GIS data layers of modeled groundwater arsenic level, representing probability of arsenic > 1µg/L, 5µg/L, and 10µg/L, respectively; *Town_Min*, *Town_Max*, and *Town_Mean* denote town-level minimum, maximum, and mean for the modeled arsenic probability values, respectively; *Smoothed_500* etc. denote the smoothed rates; e.g., *smoothed_500* denotes the town-level preterm birth rate smoothed from the *Original Rate* using a threshold that the neighborhood of smoothing must enclose at least 500 births; The bottom line contains the r values between the preterm birth rate and the town median household income in 1999; Since the term-LBW rate has very weak correlation with the income value, we did not calculate the income-adjusted r values.

Table 6

Correlation Coefficient (r) between town-level term-LBW rate for maternal age ≥ 25 and groundwater arsenic in New Hampshire.

	Original Rate	Smoothed 500	Smoothed 1000	Smoothed 1500	Smoothed 2000	Smoothed 2500	Smoothed 3000	Smoothed 3500
<i>Prob1</i>	<i>Town_Min</i>	-0.03 (0.04)	-0.02 (0.08)	-0.01 (0.12)	-0.03 (0.12)	-0.03 (0.11)	-0.02 (0.11)	-0.01 (0.11)
	<i>Town_Max</i>	-0.15 (-0.12)	-0.05 (-0.01)	-0.05 (0.00)	-0.06 (0.00)	-0.05 (0.01)	-0.03 (0.02)	-0.03 (0.02)
	<i>Town_Mean</i>	-0.06 (0.02)	-0.07 (0.02)	-0.10 (0.03)	-0.12 (0.01)	-0.11 (0.02)	-0.09 (0.03)	-0.08 (0.03)
<i>Prob5</i>	<i>Town_Min</i>	-0.03 (0.05)	0.02 (0.13)	0.02 (0.16)	-0.02 (0.13)	-0.05 (0.08)	-0.05 (0.08)	-0.05 (0.08)
	<i>Town_Max</i>	-0.06 (0.02)	-0.11 (0.00)	-0.12 (0.03)	-0.16 (-0.01)	-0.20 (-0.06)	-0.18 (-0.04)	-0.14 (-0.01)
	<i>Town_Mean</i>	-0.06 (0.04)	-0.08 (0.04)	-0.08 (0.08)	-0.11 (0.06)	-0.14 (0.03)	-0.12 (0.04)	-0.10 (0.05)
<i>Prob10</i>	<i>Town_Min</i>	-0.01 (0.05)	0.03 (0.12)	0.03 (0.14)	0.00 (0.12)	-0.03 (0.08)	-0.03 (0.08)	-0.03 (0.07)
	<i>Town_Max</i>	-0.10 (-0.02)	-0.19 (-0.09)	-0.23 (-0.10)	-0.31 (-0.17)	-0.36 (-0.24)	-0.35 (-0.22)	-0.32 (-0.20)
	<i>Town_Mean</i>	-0.05 (0.04)	-0.11 (-0.01)	-0.14 (0.00)	-0.20 (-0.05)	-0.24 (-0.10)	-0.24 (-0.10)	-0.22 (-0.09)
<i>Median household Income</i>	-0.18	-0.35	-0.37	-0.39	-0.40	-0.37	-0.33	-0.33

Notes: *Prob1*, *Prob5*, and *Prob10* denote three GIS data layers of modeled groundwater arsenic level, representing probability of arsenic $> 1\mu\text{g/L}$, $5\mu\text{g/L}$, and $10\mu\text{g/L}$, respectively; *Town_Min*, *Town_Max*, and *Town_Mean* denote town-level minimum, maximum, and mean for the modeled arsenic probability values, respectively; *Smoothed_500* etc. denote the smoothed rates; e.g., *smoothed_500* denotes the town-level preterm birth rate smoothed from the *Original Rate* using a threshold that the neighborhood of smoothing must enclose at least 500 births; The value inside the parentheses is r adjusted by town-level income (i.e., r between the arsenic value and the residual to the LBW-income linear regression); The bottom line contains the r values between the term-LBW rate and the town median household income in 1999.