Relation of Prenatal Air Pollutant and Nutritional Exposures with Biomarkers of Allergic Disease in Adolescence

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Prenatal exposures may be critical for immune system development, with consequences for allergic disease susceptibility. We examined associations of prenatal exposures (nutrient intakes and air pollutants) with allergic disease biomarkers in adolescence. We used data from 857 mother-child pairs in Project Viva, a Massachusetts-based pre-birth cohort. Outcomes of interest at follow-up (median age 12.9 years) were fractional exhaled nitric oxide (FeNO) and total serum IgE. We applied Bayesian Kernel Machine Regression analyses to estimate multivariate exposure-response functions, allowing for exposure interactions. Exposures were expressed as z-scores of log-transformed data and we report effects in % change in FeNO or IgE z-score per increase in exposure from the 25th to 75th percentile.

FeNO levels were lower with higher intakes of prenatal vitamin D (−16.15%, 95% CI: −20.38 to −2.88%), folate from foods (−3.86%, 95% CI: −8.33 to 0.83%) and n-3 PUFAs (−9.21%, 95% CI: −16.81 to −0.92%). Prenatal air pollutants were associated with higher FeNO and IgE, with the strongest associations detected for PM₂.₅ with IgE (25.6% increase, 95% CI 9.34% to 44.29%). We identified a potential synergistic interaction (p = 0.02) between vitamin E (food + supplements) and PM₂.₅; this exposure combination was associated with further increases in FeNO levels.

Allergic diseases represent a large proportion of childhood chronic disease, and constitute a major public health burden. Susceptibility to childhood allergic disease likely begins in fetal life. Fetal immune cell precursors are detectable at 3–4 weeks gestation, and begin seeding fetal tissues soon after¹. Myeloid and lymphoid cell populations continue to expand until birth, and allergic sensitization and airway inflammation in response to foreign antigen exposures evolves throughout early and later childhood, further widening the potential window of susceptibility².

Exposures encountered during prenatal life have been shown to have either adverse or, in some cases, protective associations with asthma and allergic disease in later childhood³⁴. For example, a number of studies have shown early life air pollution exposures to be risk factors for later wheeze, asthma, and reduced lung function⁵⁻⁷. Other studies have found protective associations of maternal antenatal vitamin D or antioxidant nutrient intake with later allergic or respiratory outcomes⁸⁻¹², though their findings have not always been reproducible¹³⁻¹⁶. In the antenatal and early-life periods, exposures do not occur one at a time, but as mixtures, with possible interactions amongst the exposures within those mixtures. Between-study variation in associations of individual exposures with allergy, asthma symptoms and lung function may occur in part because their effects in the context of other exposures were not explored. In addition, between-study variation in associations of exposures with allergy or

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asthma outcomes may reflect variation in subject sensitivity to those exposures and their mixtures. Recent studies have suggested sexual dimorphism in allergy and asthma that may lead to differential responses to pollutants. IgE and fractional exhaled nitric oxide (FeNO) are important intermediate biomarkers of allergic disease and allergic airway inflammation. IgE mediates the type I hypersensitivity reactions involved in pathogenesis of asthma and allergic rhinitis. FeNO is a known biomarker of eosinophilic airway inflammation, and is elevated in subjects with atopy. In adolescent children followed since the prenatal period, we examined multiple components from two prenatal exposure types, air pollution and maternal nutrient intakes, to determine their associations with IgE and FeNO measured in adolescence. We hypothesized that intake of prenatal antioxidant nutrients would mitigate the detrimental effects of prenatal air pollutants on allergic disease biomarkers measured in adolescence. We considered nutrient intakes from diet alone (as bioavailability and potential protective effects may be greatest when nutrients are absorbed from food), and then also conducted analyses using combined nutrient intakes from foods plus supplements. We also examined the potential for sexual dimorphism in the effects of mixtures of exposures on outcomes in adolescence, when the influences of sex on allergy and allergic airway inflammatory responses may be in transition.

To investigate these hypotheses, we applied Bayesian Kernel Machine Regression (BKMR), a recently developed approach for estimating the health effects of multi-exposure mixtures. Although there are other types of exposure-mixture modeling approaches, many of these have significant limitations. For example, clustering methods are often too reductionist (categorizing continuous exposures), variable selection methods (i.e. Lasso regression) typically rely on simple linear and additive models that may not adequately capture complex exposure-response relationships, and traditional hierarchical modeling techniques, while accounting for correlation amongst exposure mixture components, are typically also based on a parametric model that assumes (often unrealistic) linear and additive associations. BKMR offers distinct advantages over these other approaches, including the ability to allow for potential non-linear exposure-response relationships, to identify interactions between exposures, and to simultaneously rank the importance of hierarchical exposure groups in association with the outcome of interest. We applied BKMR methodology to data from Project Viva, a longitudinal pre-birth cohort study of prenatal exposures and chronic disease outcomes, with follow-up through adolescence.

Results
Sociodemographic and perinatal characteristics of the mothers and children enrolled in the study are shown in Table 1. Mean maternal pre-pregnancy BMI was 24.7 kg/m². Sixty six percent of the children in the study were white, 15% were black, 4% were Hispanic and 16% were other race/ethnicities. Nine percent of women reported smoking in pregnancy. Characteristics of the mother-infant pairs at enrollment were very similar to the characteristics of those with FeNO and total IgE outcome data at the early teen visit age 12 (Supplemental Table 1). Distributions of the prenatal nutrient and air pollutant exposures (25th percentile, median, 75th percentile) are shown in Table 2.

| Characteristic                       | Subjects N (%) | Geometric Mean FeNO (ppb) at age 11.9 to 16.6 N = 857 subjects | Geometric Mean Total Serum IgE (IU/ml) at age 11.9–16.6 N = 590 subjects |
|-------------------------------------|----------------|----------------------------------------------------------------|-----------------------------------------------------------------------|
| Smoking in Pregnancy                |                |                                                                |                                                                       |
| Never                               | 598 (69%)      | 18.5                                                           | 62.0                                                                  |
| Smoked during Pregnancy             | 78 (9%)        | 22.1                                                           | 89.8                                                                  |
| Former                              | 181 (21%)      | 19.0                                                           | 56.7                                                                  |
| Maternal Hay fever                  |                |                                                                |                                                                       |
| Yes                                 | 256 (30%)      | 19.7                                                           | 79.9                                                                  |
| No                                  | 600 (70%)      | 18.5                                                           | 56.9*                                                                 |
| Child's Race/Ethnicity              |                |                                                                |                                                                       |
| Black                               | 123 (15%)      | 19.8                                                           | 83.5                                                                  |
| White                               | 565 (66%)      | 18.2                                                           | 56.9                                                                  |
| Hispanic                            | 35 (4%)        | 19.5                                                           | 53.6                                                                  |
| Other                               | 134 (16%)      | 21.3                                                           | 76.4                                                                  |
| Child's sex                         |                |                                                                |                                                                       |
| Female                              | 429 (50%)      | 17.9                                                           | 58.1                                                                  |
| Male                                | 428 (50%)      | 20.0*                                                          | 68.2                                                                  |
| Season of Birth                     |                |                                                                |                                                                       |
| Winter                              | 203 (24%)      | 20.6                                                           | 65.5                                                                  |
| Spring                              | 218 (25%)      | 19.6                                                           | 66.5                                                                  |
| Summer                              | 249 (29%)      | 17.7                                                           | 67.6                                                                  |
| Fall                                | 187 (22%)      | 17.9                                                           | 51.8                                                                  |

Table 1. FeNO and Total Serum IgE levels in Adolescence by Participant Characteristics. *p value < 0.05 for t-test/ANOVA.
between Vitamin D intake from foods plus supplements and total IgE level (Table 4, Fig. 3). The point estimate a decreasing linear slope from moderate to high Vitamin D levels. In contrast, there was no clear relationship dose-response curve was not linear (Fig. 2), with a relatively flat dose-response at lower vitamin D levels, and vitamin D intake from foods showed a trend toward decreased FeNO measures in adolescence (Table 3). Overall, the FeNO outcome (Table 4), which was much larger than the vitamin E estimate from diet alone (Table 3). No − for prenatal vitamin E intake from food and supplements combined was 6.94% (95% CI 2.36 to 12.24%) for − fatty acids and folate from food sources were associated with reduced early teen FeNO levels (−9.21% (95% CI −3.86% to −0.83%) for folate. In models with IgE as the outcome, point estimates for these two nutrients were similar in magnitude and direction (Table 3). Prenatal vitamin D intake from foods showed a trend toward increased FeNO levels in adolescence (p = 0.006). There was no evidence for associations between report of smoking in pregnancy or maternal BMI and either FeNO or total serum IgE in adolescence.

For BKMR analyses, posterior inclusion probabilities were calculated for exposure groups in models for each outcome. In models for FeNO, the largest group of nutrients (Vitamin D, Vitamin C, Folate, n−3 and n−6 polyunsaturated fatty acids (PUFAs)) was consistently ranked higher than air pollutant exposures, with the second nutrient grouping (Vitamin E and β-carotene) ranked last. (Supplemental Table 2) In models for IgE, air pollutant exposures had much higher posterior inclusion probabilities than either of the individual nutrient groups, which showed similar posterior inclusion probabilities to one another. (Supplemental Table 3).

For models of FeNO, estimated dose response curves for individual exposures are shown in Fig. 1, and point estimates for percent increase in FeNO z score, with an increase in exposure from the 25th to 75th percentile, (with all other exposures are fixed at the median) are shown in Table 3. Maternal prenatal n−3 polyunsaturated fatty acids and folate from food sources were associated with reduced early teen FeNO levels (−9.21% (95% CI −16.81% to −0.92%) for n-3s and −3.86%, (95% CI −8.33% to 0.83%) for folate. In models with IgE as the outcome, point estimates for these two nutrients were similar in magnitude and direction (Table 3). Prenatal vitamin D intake from foods showed a trend toward decreased FeNO measures in adolescence (Table 3). Overall, the strongest association observed was for black carbon (BC) exposure and total IgE level (15.25% increase in total IgE was associated with an interquartile range increase in log BC exposure). In models for the FeNO outcome, the point estimate for BC was 7.39% (95% CI −1.35 to 16.91%).

We also considered nutrient exposures as combined intakes from food and supplements, if supplement intakes for a given nutrient were available in our data (Table 4). Vitamin D intake (food plus supplements) showed a strong inverse association with FeNO (−16.15%, 95% CI: −20.38 to −2.88%). The shape of the estimated dose-response curve was not linear (Fig. 2), with a relatively flat dose-response at lower vitamin D levels, and a decreasing linear slope from moderate to high Vitamin D levels. In contrast, there was no clear relationship between Vitamin D intake from foods plus supplements and total IgE level (Table 4, Fig. 3). The point estimate for prenatal vitamin E intake from food and supplements combined was 6.94% (95% CI −2.36 to 12.24%) for the FeNO outcome (Table 4), which was much larger than the vitamin E estimate from diet alone (Table 3). No association was observed for Vitamin E and total IgE levels.

In addition to modeling exposure to BC as an air pollutant, we also considered exposure to PM1.2 (Table 4, Fig. 2). PM1.2 was associated with a 0.51% increase in FeNO (95% CI −0.78 to 16.63%) and a 25.6% increase in serum total IgE (95% CI 9.34 to 44.29%). Plots of BKMR results for the effects of one exposure at various levels of another exposure showed an interaction between PM1.2 and vitamin E intake from foods plus supplements.

| Prenatal Exposure Variable | N    | 25th Percentile | Median | 75th Percentile |
|---------------------------|------|----------------|--------|-----------------|
| Vitamin C, mg/d (Foods only) | 857  | 132.6          | 168.7  | 212.5           |
| mg/d (Food + Supplements)  | 857  | 206.9          | 254.1  | 308.9           |
| Vitamin E, mg/d (Foods only) | 857  | 5.4            | 6.4    | 8.0             |
| Vitamin E, mg/d (Foods + Supplements) | 857  | 13.6           | 18.3   | 25.1            |
| Beta Carotene, μg/d (Foods only) | 857  | 2432.4         | 3521.8 | 4934.9          |
| Beta Carotene, μg/d (Foods + Supplements) | 857  | 3251.3         | 4402.6 | 5875.5          |
| Folate, μg/d (Foods only) | 857  | 287.4          | 351.3  | 421.1           |
| Folate, μg/d (Foods + Supplements) | 857  | 874.2          | 1125.1 | 1311.2          |
| Choline mg/d (Foods only) | 857  | 290.3          | 323.1  | 361.0           |
| Vitamin D, IU/d (Foods only) | 857  | 150.1          | 208.6  | 280.1           |
| Vitamin D, IU/d (Foods + Supplements) | 857  | 432.8          | 552.1  | 657.7           |
| n-6 PUFAs, gm/day (Foods only) | 857  | 10.0           | 11.7   | 13.6            |
| n-3 PUFAs gm/day (Foods only) | 857  | 0.9            | 1.1    | 1.3             |
| Black Carbon μg/m3 | 766  | 0.5            | 0.7    | 0.8             |
| PM2.5 μg/m3 | 766  | 10.6           | 11.7   | 12.9            |

Table 2. Distribution of Prenatal Exposure Variables (Nutrient Intakes from Foods and Supplements and Air Pollutant Exposures).
As shown in Fig. 3, Vitamin E is associated with larger increases in FeNO when PM$_{2.5}$ exposure is higher. This interaction was confirmed in a standard linear regression model ($p = 0.02$ for interaction term). For participants in the highest quartile of PM$_{2.5}$ exposure, Vitamin E (food plus supplements) was associated with a 8.42% increase in FeNO (95% CI $-1.18$ to $13.87$%). For participants in the highest quartile of Vitamin E (food plus supplements) exposure, PM$_{2.5}$ was associated with a 14.29% increase in FeNO levels (95% CI $0.49$ to $20.94$%). In contrast, no interaction between black carbon and vitamin E was observed in BKMR models for the FeNO outcome, and a test of the interaction term in linear regression models was not statistically significant ($p = 0.75$).

In addition to examining interactions amongst the exposures themselves, we examined potential sex interactions by conducting sex-stratified BKMR analyses and comparing sex-specific prenatal exposure-response curves (Fig. 4). Air pollutant exposure response associations were strongest for male participants. In stratified analyses, prenatal PM$_{2.5}$ exposure was associated with FeNO in males (20.47% increase, 95% CI $2.52$ to $41.56$%) but not in females ($-1.52$% decrease, 95% CI $-15.50$ to $14.79$%). The interaction between sex and PM$_{2.5}$ exposure was reproduced in a linear model ($p < 0.05$ for interaction term). We did not detect any differences in exposure response curves by sex for the total serum IgE outcome.

We also examined other exposure combinations in two additional sets of models (models for PM$_{2.5}$ exposure and food-based nutrient intakes (Supplemental Table 4) and models for black carbon exposure and Nutrients (Food + supplements) (Supplemental Table 5)). For brevity, and since main effects of exposures showed similar trends across all models, we present only the models in Table 3 (food-based nutrients and black carbon exposure) and Table 4 (nutrients from food plus supplements and PM$_{2.5}$ exposure) in the main text.

![Figure 1](https://example.com/figure1.png)

**Figure 1.** Estimated exposure response functions ($h(z)$) for prenatal nutrient intake (foods only), prenatal air pollution (Black carbon) and allergic disease outcomes in adolescence. Plots show the estimated relationship between log-transformed z scores of exposures ($z$) and outcomes; shaded areas indicate 95% credible intervals. Plots are shown for outcomes FeNO (A) and Total Serum IgE (B).

| Hierarchical Exposure Group | Nutrient Intakes from Foods Only and Air Pollution | FeNO* (% change per interquartile range increase in exposure) | Total Serum IgE* (% change per interquartile range increase in exposure) |
|-----------------------------|---------------------------------------------------|------------------------------------------------|------------------------------------------------|
|                             |                                                    | Est.     | 95% CI          | Est.     | 95% CI          |
| 1                           | Vitamin E                                          | 0.33%    | $-3.90$ to $4.75$% | 0.56%    | $-7.04$ to $8.77$% |
|                             | Beta-carotene                                     | 1.91%    | $-4.81$ to $9.11$% | 6.83%    | $-4.63$ to $19.66$% |
| 2                           | Vitamin C                                          | 0.74%    | $-2.45$ to $4.03$% | 0.33%    | $-8.52$ to $10.04$% |
|                             | Folate                                             | $-3.86$% | $-8.33$% to $0.83$% | $-4.34$% | $-12.12$ to $4.13$% |
|                             | Choline                                            | $-0.23$% | $-2.83$ to $2.43$% | 5.34%    | $-4.98$ to $16.77$% |
|                             | n-6 PUFAs                                         | $-1.00$% | $-5.84$ to $4.08$% | 6.09%    | $-5.41$ to $18.99$% |
|                             | Vitamin D                                          | $-3.24$% | $-8.39$ to $2.19$% | $-4.88$% | $-14.55$ to $5.89$% |
|                             | n-3 PUFAs                                         | $-9.21$% | $-16.81$ to $-0.92$% | $-10.51$% | $-21.26$ to $1.69$% |
| 3                           | Black Carbon                                      | 7.39%    | $-1.35$ to $16.91$% | 15.25%   | $1.92$ to $30.33$% |

*Model adjusted for maternal pre-pregnancy BMI, education, hay fever, and pregnancy smoking status (never, former, smoked during pregnancy); season of birth, sine and cosine of date at outcome measurement; and child’s race/ethnicity and sex. Estimates with $p < 0.05$ are in bold.

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In addition to examining interactions amongst the exposures themselves, we examined potential sex interactions by conducting sex-stratified BKMR analyses and comparing sex-specific prenatal exposure-response curves (Fig. 4). Air pollutant exposure response associations were strongest for male participants. In stratified analyses, prenatal PM$_{2.5}$ exposure was associated with FeNO in males (20.47% increase, 95% CI $2.52$ to $41.56$%) but not in females ($-1.52$% decrease, 95% CI $-15.50$ to $14.79$%). The interaction between sex and PM$_{2.5}$ exposure was reproduced in a linear model ($p < 0.05$ for interaction term). We did not detect any differences in exposure response curves by sex for the total serum IgE outcome.

We also examined other exposure combinations in two additional sets of models (models for PM$_{2.5}$ exposure and food-based nutrient intakes (Supplemental Table 4) and models for black carbon exposure and Nutrients (Food + supplements) (Supplemental Table 5)). For brevity, and since main effects of exposures showed similar trends across all models, we present only the models in Table 3 (food-based nutrients and black carbon exposure) and Table 4 (nutrients from food plus supplements and PM$_{2.5}$ exposure) in the main text.
Discussion

In this work, we used a state of the art, non-parametric statistical methodology, BKMR, to identify potential relationships between exposures mixtures encountered in prenatal life and biomarkers of allergic disease in adolescence. While interactions between prenatal nutrient and air pollutant exposures have been studied in the context of birth defects and childhood cognitive outcomes\cite{21,22}, this is the first study, to our knowledge, that considers how these types of prenatal exposure mixtures may relate to childhood allergic disease. We hypothesized that exposure to prenatal nutrients, particularly antioxidants, would have protective effects, and that these protective nutrients would interact with prenatal air pollutant exposures to diminish their adverse effects on childhood allergic disease. In keeping with our hypothesis, we observed that higher prenatal intakes of n-3 polyunsaturated fatty acids (a potential immunomodulatory agent), vitamin D (an antioxidant and immunomodulatory agent) and folate...
were associated with lower levels of allergic disease biomarkers. In contrast with other nutrients, total vitamin E intake in pregnancy appeared to amplify the adverse effect of prenatal exposure to PM$_{2.5}$ on FeNO measured in adolescence, suggesting a synergy between these two exposures. The effects of protective nutrients were consistent across all levels of background air pollution exposure, which suggests that there was no statistical interaction between these two exposure types. However, when considering additive effects, it is clear that protective nutrients may help reduce the overall impact of adverse exposures.

In our study, average folate intake over the 1$^{st}$ and 2$^{nd}$ trimesters (specifically from foods) showed a borderline association with lower concentrations of FeNO in adolescence, but showed no association with total serum IgE levels. Prenatal folate intake and allergic disease risk has been examined in multiple previous reports, with inconsistent results. Some reports show no association with folate$^{23,24}$, but others demonstrate associations between prenatal folate and increased risk of atopy, early childhood asthma or eczema$^{25–29}$. Differences across studies may stem from variation in prenatal exposure window (early vs. late pregnancy), underlying genetic susceptibility (maternal history of allergic disease and/or MTHFR genotype), and dose of folate. We observed a potential protective association for higher dietary prenatal folate and lower adolescent FeNO within a fairly low exposure range (150 to 1200 $\mu$g/day), which is well below the levels of supplemental folate that have been linked to increased allergic disease risk. Although the potential mechanism of action for folate on asthma/allergy phenotypes has yet to be elucidated, its role as a methyl donor suggests that effects of prenatal folate could be mediated via epigenetic modifications. However, our observation of suggestive protective associations with intake from diet alone, suggests that folate (found naturally in green leafy vegetables and other food), but not folic acid (the form found in supplements), may be responsible for the observed protective associations.

Two other prenatal nutrients with immunomodulatory potential, n-3 polyunsaturated fatty acids and vitamin D, were also associated with lower FeNO levels in our study. n-3 polyunsaturated fatty acids also showed protective effects with similar (although not statistically significant) effect sizes for total serum IgE. In Project Viva, we have previously reported a lower risk of mid childhood atopy and allergic rhinitis with more vitamin D intake in pregnancy$^{30}$. The results of the current study extend these findings by examining a biomarker of allergic disease (FeNO) at a later time point. In addition to our Project Viva findings, results from clinical trials also show reduced childhood allergic disease risk with higher intakes of these prenatal nutrients. A recent randomized controlled trial published in the New England Journal of Medicine showed lower rates of persistent wheeze or asthma in children whose mothers were supplemented with 2.4 g n–3 polyunsaturated fatty acids (in fish oil) vs. olive oil starting at 24 weeks gestation$^{31}$. Secondary analyses of the VDAART randomized trial demonstrate lower odds of

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**Figure 3.** Prenatal Vitamin E Intake (Foods + Supplements) may interact with PM$_{2.5}$ Exposure to increase FeNO levels in adolescence. In subjects with elevated prenatal exposure to PM$_{2.5}$, prenatal Vitamin E shows a stronger association with FeNO in adolescence.

**Figure 4.** Estimated response function ($h(z)$) for prenatal exposure to PM$_{2.5}$ and FeNO in adolescence vary by sex. Prenatal PM$_{2.5}$ exposure ($z$) was associated with increased FeNO in males (A), but not females (B).
appeared to be associated with higher FeNO in adolescence, particularly in the context of higher PM 2.5 levels. We also tested 1st and 2nd trimester exposures to nutrients and air pollutants from the prenatal period through early adolescence are likely important, many of which may have import for health effects. Lastly, we considered only prenatal exposures. Cumulative intakes are likely to be reasonable approximations of dietary intakes in the 3rd trimester. We assessed air pollution exposure by PM 2.5, and black carbon (a component of PM2.5), but did not interrogate additional PM components, exposure to 3rd trimester air pollutants did not directly overlap with our nutrient assessments from the 1st and 2nd trimesters. It is possible that we would have observed more interaction effects if we had data on 3rd trimester exposure to 3rd trimester air pollutants did not directly overlap with our nutrient assessments from the 1st and 2nd trimesters. It is possible that we would have observed more interaction effects if we had data on 3rd trimester nutrient intakes were concurrent with air pollutant exposures; however the 1st and 2nd trimester nutrient intakes are likely to be reasonable approximations of dietary intakes in the 3rd trimester. We assessed air pollution exposure by PM2.5 and black carbon (a component of PM2.5), but did not interrogate additional PM components, many of which may have import for health effects. Lastly, we considered only prenatal exposures. Cumulative exposures to nutrients and air pollutants from the prenatal period through early adolescence are likely important, but were beyond the scope of this analysis.

In summary, we found that greater maternal intakes of vitamin D, folate, and n-3 fatty acids were associated with less airway inflammation in adolescence, while other nutrients (vitamin E), and air pollutants were associated with higher inflammation. Prenatal exposure interactions may play an important role, as evidenced by the potential synergy between vitamin E and PM2.5. We also observed potential protective effects of prenatal nutrients (n-3 PUFAs) and adverse associations with prenatal air pollution exposure for a second allergic disease biomarker (total IgE) in adolescence. Whenever possible, exposure interactions in prenatal life should be examined in studies of chronic disease outcomes in childhood. Understanding how prenatal exposures, particularly in the context of mixtures, influence allergic disease in childhood may ultimately inform the design of prenatal intervention trials to reduce the incidence of these disorders.
Methods
This study involved women and their children enrolled in Project Viva, an ongoing longitudinal pre-birth cohort study. The design of this cohort study has been described in detail elsewhere. The study was approved by the Brigham and Women’s Hospital Institutional Review Board and the Harvard Pilgrim Health Care Institute Institutional Review Board. All methods were performed in accordance with the guidelines and regulations of these institutional review boards. Mothers provided informed consent at enrollment and for their child during follow-up visits, and children gave verbal assent. Participants were recruited into Project Viva at 8 offices of Atrius Harvard Vanguard Medical Associates in Eastern Massachusetts between 1999 and 2002. Exclusion criteria were multiple gestation, inability to answer questions in English, plans to move out of the study area before delivery of the infant, and gestational age ≥ 22 weeks at the time of presentation for prenatal care. We saw mothers during the first and second trimesters of pregnancy, and both mothers and children at delivery and periodic postnatal research visits, most recently when children reached adolescence (the “Early Teen” visit at median 12.9 years, range 11.9 to 16.6 years). At this visit, we measured respiratory function and collected blood.

Of the 2,128 women who delivered a live infant, we excluded 270 who were missing all prenatal exposure data (nutrient intakes and air pollution exposure estimates), and 22 who were missing relevant covariates. Of the 1,836 mother-infant pairs with available exposures and covariate measures, 857 (47%) had child’s FeNO and 590 (32%) had child’s total IgE measured at the Early Teen visit. Comparison of the 2,128 participants to the 1,836 mother-infant pairs with measured exposures and covariates showed some differences. For example, maternal pre-pregnancy body mass index (BMI) was lower for participants with exposure measures (mean 24.6 vs. 26.5 kg/m², t-test p value = 0.0001). A larger percentage of participants with prenatal exposures measures were white (67 vs. 39%), and highly educated (68% with a college degree vs 36%), but the two groups did not differ on maternal history of hay fever. Comparisons between the 1,836 participants with exposure and covariate data and those with FeNO and total IgE outcome data in adolescence are shown in Supplemental Table 1.

Prenatal Dietary Nutrients and Supplement Intakes. During both the first and second trimesters, data on prenatal nutrient intakes were gathered using semi-quantitative food frequency questionnaires (FFQs) validated specifically for use in pregnancy. We used the average of 1st and 2nd trimester FFQ derived nutrient intakes for our analyses. Use of these questionnaires in the Project Viva cohort have been described previously. For first trimester data, the time referent was “during this pregnancy” (i.e. from the date of the last menstrual period until the assessment at a median of 9.9 weeks gestation at enrollment). The time referent for second trimester intakes collected by FFQ at the second visit (at 26–28 weeks gestation) was “during the past 3 months”. We did not administer a full FFQ that covered diet during late pregnancy, but we have found that diet was fairly stable across the first and second trimesters. In addition to the FFQ, we also gathered data on vitamin and supplement intakes, including information on dose, duration of use, brand or type of multivitamin, prescribed prenatal vitamins and supplement use by the mother during both the first and second trimesters. We estimated individual nutrient intakes using the Harvard nutrient-composition database, which contains food composition values from the US Department of Agriculture, supplemented by other data sources. We used mean nutrient intakes from the first and second trimesters as the exposures in the analyses, adjusted for total energy intake by the nutrient residual method. If a participant completed only one FFQ, we used that value for the exposure variable.

Prenatal Air Pollutant Exposures. We estimated third trimester air pollutant exposures of interest, using validated spatio-temporal regression models. Methodology for generating daily estimates of these exposures has been published elsewhere. Briefly, daily residence-specific BC estimates were generated from a spatiotemporal regression model that contained land-use predictors (e.g., cumulative traffic-density within 100 m of a given location), meteorological terms, and a smooth term of longitude and latitude. This model was applied separately for predictions in the warm (May-October) and cold (November-April) seasons. For BC estimates obtained using this spatiotemporal land-use regression model, the mean “out-of-sample” R² was 0.73. For PM2.5 exposure estimates, we used a spatiotemporal prediction model that uses fine-scale spatial information at 100 m resolution daily satellite remote sensing data on aerosol optical depth (AOD) from the NASA MODIS satellite, land use terms, and meteorological factors. The models first calibrate the remote sensing AOD data to observed PM2.5 ground monitoring data while accounting for land use and meteorological factors. A second stage uses generalized additive models to fill in missing AOD data, due to cloud cover, snow, or other factors, using regional measured PM2.5 AOD values in neighboring cells, and land use terms. A final stage uses fine-scale spatial information at 100 m resolution to account for very local traffic particle emissions. Daily PM2.5 exposure estimates at each residence yielded mean “out-of-sample” R² of 0.83 for days having AOD remote sensing data and 0.81 for days without. Participants reported their residential address at enrollment and updated it at the end of the second trimester and shortly after birth. All air pollution exposure estimates account for any moves during pregnancy. Third trimester exposure estimates were calculated by averaging daily exposures from the 188th day after LMP (last menstrual period) to the day before birth. We chose to model 3rd trimester air pollutant exposures, as 1) the third trimester is a very active period of lung development and 2) the 3rd trimester exposure data were most complete for mothers who enrolled at the beginning of the study, which began in 1999.

Fractional Exhaled Nitric Oxide (FeNO) Measurement. Nitric oxide (NO) is a mediator involved in chronic inflammatory diseases and Th2-mediated immune responses. Measurement NO in exhaled air, as fractional exhaled nitric oxide (FeNO), is a non-invasive biomarker of airway inflammation that correlates with airway eosinophilia. We measured FeNO levels with a portable electrochemical device (NIOX MINO; Aerocrine
AB, Stockholm, Sweden), which is in agreement with published procedures for FeNO measurement. Before FeNO measurement, participants breathed in through a nitric oxide scrubbing filter and exhaled out into the room air twice before inhaling a third time through the filter and exhaling into the FeNO analyzer. Exhaled FeNO measurements were conducted without a nose clip at a flow rate of 50 mL/s. (The last 3 seconds of the exhalation was used for FeNO measurement to ensure quantification of lower rather than upper airway FeNO). In accordance with American Thoracic Society guidelines, we did not use nose clips because blocking the nasal passages can cause exhaled nitric oxide to accumulate in the nose and potentially leak into the exhaled air stream through the posterior nasopharynx. This procedure was performed twice, so that each subject had two FeNO measurements. The mean of these two measurements was used in data analysis.

Total IgE Measurement. Of the 1,836 children with prenatal exposure assessment for air pollutants and nutrients, 706 had blood drawn at school age for additional studies, of whom 590 had sufficient sample to measure serum total IgE by using ImmunoCAP (Phadia, Uppsala, Sweden).

Covariates. At enrollment mothers completed questionnaires reporting their education, history of atopy, smoking habits, and pre-pregnancy height and weight, from which we calculated BMI. On postpartum questionnaires, mothers reported child race/ethnicity. We obtained child sex and date of birth from hospital birth records.

Statistical Analysis. We applied Bayesian Kernel Machine Regression (BKMR), which estimates the multivariable exposure-response function of the prenatal nutrients and air pollutant exposures with each of our two outcomes (FeNO and IgE levels). BKMR is a flexible modeling approach that allows for a non-linear and non-additive (e.g., interactive) exposure-response relationships, and which simultaneously provides measures of variable importance (posterior inclusion probabilities). Using BKMR, it is possible to estimate a flexible exposure-response function (denoted $h(\beta)$). For each BKMR model, we performed 10,000 iterations. We applied the hierarchical variable approach in order to rank the importance of exposure groups in association with FeNO and total serum IgE measured at the early teen visit. Prenatal maternal exposures of interest were residence-specific air pollutants (3rd trimester BC or PM$_{2.5}$) and nutrient intakes (1st and 2nd trimester averaged energy-adjusted dietary intakes of vitamins D, C, and E, β-carotene, folate, choline, and n-3 and n-6 polysaturated fatty acids). We conducted secondary analyses with nutrient intakes derived from supplements and food combined. We modelled all nutrients in the same exposure group except vitamin E and β-Carotene, which demonstrated associations with FeNO and total IgE in the direction opposite to that observed for other nutrients. In our BKMR model specification, we grouped the exposures into 3 different groups, vitamin E and β-Carotene (group 1), all of the other nutrients (group 2), and the air pollutant (either PM$_{2.5}$ or BC) exposure (group 3) and conducted hierarchical variable selection on these groups of exposures. In particular, under the hierarchical variable selection formulation, BKMR simultaneously identifies (1) whether there are particular groups of exposures that are associated with the outcome, and (2) which exposures within the group may be driving the association between that group of the pollutants with the outcome by calculating group-specific importance measures, which are referred to as ‘posterior inclusion probabilities’. We used posterior inclusion probabilities from the BKMR model, which is an importance score, to rank the strength of association between each exposure group and the outcome. Models were adjusted for maternal education, hay fever, pre-pregnancy BMI, and smoking in pregnancy, and child sex, race/ethnicity, season of birth, and size and cosine of time of outcome measure to adjust for confounding by seasonal trends in both the outcome and the exposure. Exposure covariates were expressed as $z$-scores of log-transformed exposures and we report effects in % change in FeNO or IgE $z$-per interquartile range (25th to 75th percentile) increase in exposure. Since BC is a component of PM$_{2.5}$, we modeled the two air pollutant exposures separately. In all, four combinations of variable types were run in BKMR analyses: Model 1: Nutrients (foods only) + BC air pollutant; Model 2. Nutrients (foods only) + PM$_{2.5}$ air pollution; Model 3: Nutrients (foods + supplements) + BC, Model 4: Nutrients (foods + supplements) + PM$_{2.5}$ air pollution. Models 1 and 4 are presented in the main text for the outcomes FeNO and serum total IgE. Results of models 2 and 3 are shown in the supplement. We confirmed the main effects and interactions estimated in the BKMR analyses with standard parametric regression models with interaction terms. We refer to associations as statistically significant if 95% posterior credible intervals exclude the null.

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J.E.S performed analyses, interpreted data and wrote manuscript. B.A.C. helped with air pollution exposure assessment, data interpretation, and also guided statistical analyses with BKMR methodology. J.B. guided statistical analyses with BKMR methodology. S.L.R. assisted with data management, data interpretation and manuscript editing. J.S. helped with data interpretation and manuscript editing. I.K. helped with air pollution exposure assessment, data interpretation and manuscript editing. H.G. helped with data management, data interpretation and manuscript editing. K.M.S. helped with data interpretation and manuscript editing. A.A.L. helped with data interpretation and manuscript editing. P.K. helped with air pollution exposure assessment, data interpretation and manuscript editing. A.R.G. conceived of the study design and assisted with data interpretation and manuscript writing. E.O. conceived of the study design and assisted with data interpretation and manuscript writing. All authors read, approved and commented on the final manuscript.

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