Does Living Near a Superfund Site Contribute to Higher Polychlorinated Biphenyl (PCB) Exposure?

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Does Living Near a Superfund Site Contribute to Higher Polychlorinated Biphenyl (PCB) Exposure?


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We assessed determinants of cord serum polychlorinated biphenyl (PCB) levels among 720 infants born between 1993 and 1998 to mothers living near a PCB-contaminated Superfund site in Massachusetts, measuring the sum of 51 PCB congeners (ΣPCB) and ascertaining maternal address, diet, sociodemographics, and exposure risk factors. Addresses were geocoded to obtain distance to the Superfund site and neighborhood characteristics. We modeled log10(ΣPCB) as a function of potential individual and neighborhood risk factors, mapping model residuals to assess spatial correlates of PCB exposure. Similar analyses were performed for light (mono–tetra) and heavy (penta–deca) PCBs to assess potential differences in exposure pathways as a function of relative volatility. PCB-118 (relatively prevalent in site sediments and cord serum) was assessed separately. The geometric mean of ΣPCB levels was 0.40 (range, 0.068–18.14) ng/g serum. Maternal age and birthplace were the strongest predictors of PCB levels. Maternal consumption of organ meat and local dairy products was associated with higher and smoking and previous lactation with lower PCB levels. Infants born later in the study had lower PCB levels, likely due to temporal declines in exposure and site remediation in 1994–1995. No association was found between PCB levels and residential distance from the Superfund site. Similar results were found with light and heavy PCBs and PCB-118. Previously reported demographic (age) and other (lactation, smoking, diet) correlates of PCB exposure, as well as local factors (consumption of local dairy products and Superfund site dredging) but not residential proximity to the site, were important determinants of cord serum PCB levels in the study community. Key words: exposure pathways, geographic information systems, hazardous waste site, newborn, PCBs, polychlorinated biphenyls, remediation, Superfund. Environ Health Perspect 114:1092–1098 (2006). doi:10.1289/ehp.8827 available via http://dx.doi.org/ [Online 23 January 2006]
Cord blood PCB levels. Cord blood samples were obtained at birth in Vacutainer tubes and centrifuged, and the serum fraction was removed. The serum was stored in solvent-rinsed glass vials with Teflon-lined caps at –20°C until extraction. Analyses were performed by the Harvard School of Public Health Organic Chemistry Laboratory. Cord blood analytic methods and quality control procedures are described elsewhere (Korrick et al. 2000). Briefly, 51 individual PCB congeners were measured using liquid–liquid extraction and extract analysis by capillary column gas chromatography with electron capture detection. Confirmatory analyses were done with microelectron capture detection and a capillary column of different polarity. Serum lipids were not measured because of insufficient sample volume. PCB concentrations were reported as the sum of 51 congeners (ΣPCB) in units of nanograms of analyte per gram of serum. We also grouped PCB levels into light PCBs (sum of 14 mono- to tetrachlorinated biphenyls) and heavy PCBs (sum of 37 pentato decachlorinated biphenyls) according to their elution order and relative volatility (Cullen et al. 1996). These two groups were chosen a priori based on the hypothesis that PCB exposure pathways may vary by their relative volatility.

The 51 congeners were chosen based on their toxicity, persistence in the environment or human samples, and presence in New Bedford environmental samples; these included a subset of mono-ortho dioxin-like PCBs (congeners 105, 118, 156, 167, and 189). The dioxin TEQ concentration for the dioxin-like PCBs was calculated (Van den Berg et al. 1998) and expressed in parts per trillion (ppt) lipid, assuming 0.17% lipid for cord serum based on our laboratory’s data and published values (Alshul LM, personal communication; Denkins et al. 2000). PCB-118 was chosen a priori for individual assessment. It was prevalent in harbor sediments consistent with the predominant Aroclors used by the area’s industries (Brown and Wagner 1990; Weaver 1984). In addition, it was disproportionately prevalent in our serum samples; cord serum levels of PCB-118 were comparable with levels observed in other population-based surveys (Korrick et al. 2000) despite overall PCB levels being substantially lower than most other populations (Longnecker et al. 2003).

Dietary assessment. Mothers completed a semiquantitative food-frequency questionnaire during a home evaluation of the child at age 2 weeks. The mothers reported diet histories before and during pregnancy. Twenty-four items from the food-frequency questionnaire were collapsed into six groups: meat (including organ meat), poultry, dairy, eggs, grains, and fish. We further considered fish in four subcategories: tuna; dark-meat fish (mackerel, blue fish, salmon, sardines, and swordfish); other fish (including catfish), and shellfish. In addition, mother’s self-reported consumption of locally grown produce, dairy products (including eggs), meat (including chicken), fish, game, and wine were determined as binary (yes/no) variables.

Occupation, gardening, and other potentially PCB-exposure-related activities. Mothers’ potential occupational PCB exposure (including working with paints, sealants, caulking compounds, and lubricants), gardening, and other potentially PCB-exposure–related activities (including use of pesticides and fertilizers) were determined by interviewer-administered questionnaire at the 2-week evaluation. Total person-years of exposure were calculated separately for occupation, gardening, and other. Potential exposures were reported as the sum of years from self-reports of engagement in these activities for at least 1 day per week. For each exposure pathway, we divided person-years of potential exposure into three categories: zero and below and above the 75th percentile of nonzero values.

Other risk factors. We determined maternal age, birthplace, race, education, marital status, reproductive history, pregnancy smoking and alcohol consumption, residential history, household income, and infant’s race and sex from the 2-week questionnaire and maternal and infant medical records.

Geographic information systems. Home address for the duration of the mother’s pregnancy was geocoded by Mapping Analytics (Rochester, NY), a commercial geocoding firm previously shown to have good (96%) accuracy (Krieger et al. 2001). We used the geocoded residence location for mapping, calculating distance from the Superfund site, and retrieving Census block group data.

A map of New Bedford Harbor PCB levels (U.S. EPA 2001) was aligned to the Massachusetts town boundaries (MassGIS...
2002; scale 1:25,000 meter units) using ArcGIS (ESRI Inc., Redlands, CA) to estimate the latitude and longitude of the harbor hot spot. Residential distance (in miles) from the hot spot was used as an index of potential site-related PCB exposure.

Indoor PCB sources include pre-1977 sealants, electrical appliances, and light fixtures (Balfanz et al. 1993; Vorhees et al. 1997). We did not have information about individual home characteristics, but as a proxy, we calculated the fraction of houses built between 1940 and 1979 compared with the total number of houses built through 1990, using 1990 Census block group data.

We constructed neighborhood socioeconomic indices based on 1990 Census block group data (Krieger et al. 2003): a) crowding—percentage of households with more than one person per room; b) poverty—percentage of persons below the federally defined poverty line ($12,647 for a family of four in 1989); c) low income—percentage of households with income less than 60% of the U.S. median household income ($18,000); d) median household income; e) high education—percentage of persons, 25 or more years of age, with at least 4 years of college; and f) low education—percentage of persons, 25 or more years of age, with less than a 12th grade education.

Statistical analysis. The cord serum PCB levels were highly positively skewed and were log10 transformed for linear regression analyses. Univariate and bivariate associations were explored. Associations between log PCB levels and continuous covariates were assessed using scatter plot smoothing (Venables and Ripley 1997) to examine any nonlinear relationships.

Potential exposure risk factors were divided into those associated with exposure pathways—dietary, inhalation, and dermal exposure sources—and those related to individual characteristics. A set of core individual characteristics was included in each exposure pathway analysis: maternal age and birthplace, smoking during pregnancy, previous lactation, child’s date of birth and sex, dredging period, and household income. Individual socioeconomic indicators (maternal education and race) were also included in models assessing neighborhood socioeconomic indicators and PCB levels. Multivariate models for log PCB included the core individual characteristics and exposure pathway covariates significant (p < 0.10) in at least one of the individual pathway models for at least one of the PCB measures (ΣPCB, heavy PCBs, light PCBs and PCB-118). Regression results are reported as the relative (percent) increase in PCB level associated with each predictor, calculated as the antilog of the regression coefficient and 95% confidence intervals.

PCB levels were mapped and a smoothed surface was fitted by kriging (Cressie 1991) using ArcGIS Geostatistical Analyst (ESRI Inc.). We estimated the surface by an inverse-distance weighted average of 25 neighboring points chosen on the basis of a small-prediction mean square error and a reasonable area to detect local spatial variability. We restricted this mapping to residences within a 5-mile radius of the hot spot. Similar mapping was performed for multivariate model residuals to provide information on any unmeasured spatial correlates of PCB exposure. To protect the confidentiality of participants, each residence location was offset by a random amount generated from a normal distribution with mean zero and standard deviation (SD) equal to 1% of the SD of residence latitudes and longitudes.

Generalized additive models (Hastie and Tibshirani 1990) were fit in S-Plus (version 3.4; Insightful Corp., Seattle, WA) to assess temporal variability in PCB levels. The span parameter with the lowest Akaike information criterion (AIC) for each PCB measure was chosen.

Table 1. Maternal, infant, and household characteristics and associated PCB levels (unadjusted geometric means, ng/g serum) among 720 mother–infant pairs in the Greater New Bedford area.

<table>
<thead>
<tr>
<th>Characteristic</th>
<th>Total</th>
<th>ΣPCB</th>
<th>Heavy PCBs</th>
<th>Light PCBs</th>
<th>PCB-118</th>
</tr>
</thead>
<tbody>
<tr>
<td>n (%)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Maternal age (years)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>&lt; 20</td>
<td>104 (15)</td>
<td>0.31</td>
<td>0.23</td>
<td>0.061</td>
<td>0.026</td>
</tr>
<tr>
<td>20–24</td>
<td>224 (31)</td>
<td>0.30</td>
<td>0.24</td>
<td>0.054</td>
<td>0.027</td>
</tr>
<tr>
<td>25–29</td>
<td>204 (28)</td>
<td>0.42</td>
<td>0.25</td>
<td>0.061</td>
<td>0.040</td>
</tr>
<tr>
<td>30–34</td>
<td>130 (19)</td>
<td>0.58</td>
<td>0.49</td>
<td>0.079</td>
<td>0.048</td>
</tr>
<tr>
<td>≥ 35</td>
<td>50 (7)</td>
<td>0.73</td>
<td>0.63</td>
<td>0.083</td>
<td>0.051</td>
</tr>
<tr>
<td>p-Trend</td>
<td>p &lt; 0.0001</td>
<td>p &lt; 0.0001</td>
<td>p &lt; 0.0001</td>
<td>p &lt; 0.0001</td>
<td></td>
</tr>
<tr>
<td>Infant’s date of birth</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Before dredging</td>
<td>140 (19)</td>
<td>0.53</td>
<td>0.41</td>
<td>0.100</td>
<td>0.038</td>
</tr>
<tr>
<td>During dredging</td>
<td>216 (30)</td>
<td>0.46</td>
<td>0.38</td>
<td>0.071</td>
<td>0.045</td>
</tr>
<tr>
<td>After dredging</td>
<td>364 (51)</td>
<td>0.33</td>
<td>0.27</td>
<td>0.049</td>
<td>0.029</td>
</tr>
<tr>
<td>p-Heterogeneity</td>
<td>p &lt; 0.0001</td>
<td>p &lt; 0.0001</td>
<td>p &lt; 0.0001</td>
<td>p &lt; 0.0001</td>
<td></td>
</tr>
<tr>
<td>Maternal race</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Black/African American</td>
<td>24 (4)</td>
<td>0.33</td>
<td>0.27</td>
<td>0.059</td>
<td>0.027</td>
</tr>
<tr>
<td>Latino</td>
<td>46 (8)</td>
<td>0.37</td>
<td>0.30</td>
<td>0.063</td>
<td>0.037</td>
</tr>
<tr>
<td>Othera</td>
<td>52 (9)</td>
<td>0.48</td>
<td>0.39</td>
<td>0.077</td>
<td>0.045</td>
</tr>
<tr>
<td>Non-Hispanic white</td>
<td>460 (79)</td>
<td>0.40</td>
<td>0.33</td>
<td>0.060</td>
<td>0.035</td>
</tr>
<tr>
<td>p-Heterogeneity</td>
<td>p = 0.13</td>
<td>p = 0.13</td>
<td>p = 0.13</td>
<td>p = 0.10</td>
<td></td>
</tr>
<tr>
<td>Maternal education</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>High school or less</td>
<td>335 (58)</td>
<td>0.38</td>
<td>0.31</td>
<td>0.061</td>
<td>0.035</td>
</tr>
<tr>
<td>Some college or higher</td>
<td>247 (42)</td>
<td>0.42</td>
<td>0.34</td>
<td>0.062</td>
<td>0.037</td>
</tr>
<tr>
<td>p-Value</td>
<td>p = 0.18</td>
<td>p = 0.14</td>
<td>p = 0.71</td>
<td>p = 0.36</td>
<td></td>
</tr>
<tr>
<td>Previous lactation</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>0–6 months</td>
<td>514 (88)</td>
<td>0.39</td>
<td>0.32</td>
<td>0.061</td>
<td>0.036</td>
</tr>
<tr>
<td>&gt; 6 months</td>
<td>68 (12)</td>
<td>0.44</td>
<td>0.37</td>
<td>0.060</td>
<td>0.035</td>
</tr>
<tr>
<td>p-Value</td>
<td>p = 0.19</td>
<td>p = 0.14</td>
<td>p = 0.86</td>
<td>p = 0.91</td>
<td></td>
</tr>
<tr>
<td>Maternal smoking during pregnancy</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>No</td>
<td>392 (67)</td>
<td>0.43</td>
<td>0.35</td>
<td>0.032</td>
<td>0.042</td>
</tr>
<tr>
<td>Yes</td>
<td>190 (33)</td>
<td>0.34</td>
<td>0.27</td>
<td>0.054</td>
<td>0.026</td>
</tr>
<tr>
<td>p-Value</td>
<td>p = 0.0001</td>
<td>p = 0.0001</td>
<td>p = 0.002</td>
<td>p &lt; 0.0001</td>
<td></td>
</tr>
<tr>
<td>Annual household income</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>&lt; $40,000</td>
<td>383 (70)</td>
<td>0.39</td>
<td>0.32</td>
<td>0.061</td>
<td>0.035</td>
</tr>
<tr>
<td>≥ $40,000</td>
<td>184 (30)</td>
<td>0.46</td>
<td>0.38</td>
<td>0.066</td>
<td>0.042</td>
</tr>
<tr>
<td>p-Value</td>
<td>p = 0.01</td>
<td>p = 0.01</td>
<td>p = 0.27</td>
<td>p = 0.02</td>
<td></td>
</tr>
</tbody>
</table>

*Other race includes Asian and Native American and nonwhite Cape Verdean.

Results

Cord serum PCB levels had geometric means (SDs) as follows: ΣPCB, 0.40 (2.02) ng/g with a range of 0.068–18.14 ng/g; heavy PCBs, 0.33 (2.09) ng/g with a range of 0.035–11.91 ng/g; light PCBs, 0.063 (2.12) ng/g with a range of 0.0074–6.23 ng/g; PCB-118, 0.035 (2.37) ng/g with a range of 0–2.05 ng/g; and dioxin-like PCB TEQs, 4.40 (2.39) ppt lipid, with a range of 0–151.5 ppt lipid.

Maternal, infant, and household characteristics are shown in Table 1. Twenty percent of mothers were born outside of the United States (14% from Portugal, the Azores, or Cape Verde). Most (58%) had an educational level of high school or less, and 70% had an annual household income of < $40,000; 75% of the study population resided within 3.9 miles of the harbor hot spot. Cord serum PCB levels had geometric means (range of 0–151.5 ppt lipid). Linear regression models were fit in SAS (version 8.2; SAS Institute Inc., Cary, NC).

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the hot spot. Half of the infants were born after the harbor was dredged (October 1995 and later).

Maternal age at the infant’s birth was strongly associated with cord serum PCB levels, which declined over time, with additional declines after the harbor dredging was completed (Table 1, Figure 2). Mothers who were born in Portugal, the Azores, or Cape Verde and female infants had significantly higher cord serum PCB levels (Table 1). After adjustment for maternal age, prior lactation and higher household income were associated with lower cord serum PCB levels. Maternal smoking during pregnancy was also associated with lower PCB levels (Table 1). These parameters were defined as core covariates and included in subsequent analyses. Maternal marital status, alcohol consumption during pregnancy, and infant race were not associated with serum PCB levels.

PCB associations with maternal diet before and during pregnancy were essentially the same. We report the results of analyses assessing diet during pregnancy. Maternal intake of organ meats (liver, tripe, kidney, bone marrow) was significantly associated with higher PCB levels ($p < 0.05$) for PCB, light PCBs, and PCB-118 after adjustment for the base model covariates ($p = 0.05$ for heavy PCBs) (Table 2).

Consumption of local dairy products (including eggs) was associated with significantly higher levels of ΣPCB and heavy PCBs. Consumption of dark fish was positively associated with PCB levels, but this association was only marginally significant for light PCBs and PCB-118 (Table 2).

Mothers who were long-term gardeners had infants with lower heavy PCB and PCB-118 levels than infants of mothers who did not garden; however, this association was based on a very small sample size ($n = 9$) and therefore was not included in our final multivariate model. Otherwise, we found no consistent association of cord serum PCB levels with PCB-related occupations or activities, distance of residence from the hot spot, or age of homes in the child’s neighborhood (Table 3). Although there was a tendency for infants born to mothers living in poor or low-income neighborhoods to have higher light PCB levels than those born to mothers living in other neighborhoods, these associations were not significant (Table 4).

We constructed multivariate models including core covariates and significant covariates from the pathway analyses (Table 5). Maternal age and birthplace (in Portugal, the Azores, or Cape Verde) remained the strongest predictors of cord serum PCB levels ($p < 0.001$). In addition, infants born late in the study had significantly lower PCB levels than infants born early in the study (Table 5). Even with adjustment for infant birth date, infants born after dredging had significantly lower light PCB and PCB-118 levels, with near significance for ΣPCB levels (Table 5). Covariate-adjusted smoothed plots of ΣPCB, heavy PCB, light PCB, and PCB-118 levels by infant date of birth corroborate the apparent independent dredging effect (Figure 2). Mother’s prior lactation and smoking during pregnancy were significantly associated with lower PCB levels, and maternal consumption of organ meat and locally produced dairy were associated significantly with higher PCB levels (Table 5).

Maps of unadjusted log$_{10}(Σ$PCB) levels (Figure 1A) and log$_{10}(Σ$PCB) residuals from the multivariate adjusted model (Figure 1B) showed spatial variability in PCB levels but no relationship to proximity of residence to the PCB hot spot. Similar results were found with heavy and light PCB levels.

Results of pathway analyses for dioxin-like PCB TEQs were similar to those of the four other PCB measures. Significant predictors of higher PCB TEQ concentrations included older maternal age; maternal birth in Portugal, the Azores, or Cape Verde; and consumption of red meat during pregnancy. Mother’s previous

---

**Table 2. Percent change in newborn cord serum PCB levels associated with maternal dietary exposure.**

<table>
<thead>
<tr>
<th>General diet</th>
<th>ΣPCB</th>
<th>Heavy PCBs</th>
<th>Light PCBs</th>
<th>PCB-118</th>
</tr>
</thead>
<tbody>
<tr>
<td>No. with given dietary exposure$^a$</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Red meat (&gt; 2/week)</td>
<td>394</td>
<td>7</td>
<td>6</td>
<td>13$^*$</td>
</tr>
<tr>
<td>Organ meat (&gt; 1/month)</td>
<td>49</td>
<td>23$^{**}$</td>
<td>20$^*$</td>
<td>32$^{**}$</td>
</tr>
<tr>
<td>Chicken/turkey (&gt; 1/month)</td>
<td>511</td>
<td>2</td>
<td>4</td>
<td>-1</td>
</tr>
<tr>
<td>Dairy (&gt; 5/day)</td>
<td>257</td>
<td>-5</td>
<td>-6</td>
<td>-3</td>
</tr>
<tr>
<td>Eggs (&gt; 2/week)</td>
<td>296</td>
<td>-6</td>
<td>-5</td>
<td>-5</td>
</tr>
<tr>
<td>Dark fish (&gt; 1/month)</td>
<td>120</td>
<td>10</td>
<td>9</td>
<td>15$^*$</td>
</tr>
<tr>
<td>Other fish (&gt; 2/week)</td>
<td>112</td>
<td>1</td>
<td>2</td>
<td>-3</td>
</tr>
<tr>
<td>Tuna (&gt; 2/week)</td>
<td>197</td>
<td>-1</td>
<td>-2</td>
<td>1</td>
</tr>
<tr>
<td>Shellfish (&gt; 2/week)</td>
<td>249</td>
<td>-1</td>
<td>-1</td>
<td>-4</td>
</tr>
<tr>
<td>Grain (&gt; 3/day)</td>
<td>264</td>
<td>-3</td>
<td>-3</td>
<td>-2</td>
</tr>
<tr>
<td>Locally grown food (yes)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Fish</td>
<td>54</td>
<td>6</td>
<td>8</td>
<td>2</td>
</tr>
<tr>
<td>Wine</td>
<td>27</td>
<td>6</td>
<td>7</td>
<td>8</td>
</tr>
<tr>
<td>Produce</td>
<td>398</td>
<td>-6</td>
<td>-7</td>
<td>-5</td>
</tr>
<tr>
<td>Dairy</td>
<td>53</td>
<td>27$^{**}$</td>
<td>31$^{**}$</td>
<td>9</td>
</tr>
<tr>
<td>Game</td>
<td>20</td>
<td>-3</td>
<td>-4</td>
<td>-3</td>
</tr>
<tr>
<td>Meat</td>
<td>25</td>
<td>-16</td>
<td>-21</td>
<td>-4</td>
</tr>
</tbody>
</table>

$^a$Reflects percent change in a multiplicative scale, obtained by exponentiating the regression coefficient in the log-transformed PCB model. $^b$Adjusted for base variables (maternal age, birthplace, smoking during pregnancy, previous lactation, household income, child’s date of birth and sex, and dredging period). $^c$Represents number of individuals in each food consumption group from a total sample of 531 with nonmissing diet data. $^d$Reflects percent change in a multiplicative scale, obtained by exponentiating the regression coefficient in the log-transformed PCB model. $^e$Adjusted for base variables (maternal age, birthplace, smoking during pregnancy, previous lactation, household income, child’s date of birth and sex, and dredging period). $^f$Represents number of individuals in each food consumption group from a total sample of 531 with nonmissing diet data. $^g$Reflects percent change in a multiplicative scale, obtained by exponentiating the regression coefficient in the log-transformed PCB model. $^h$Adjusted for base variables (maternal age, birthplace, smoking during pregnancy, previous lactation, household income, child’s date of birth and sex, and dredging period). $^i$Represents number of individuals in each food consumption group from a total sample of 531 with nonmissing diet data. $^j$Reflects percent change in a multiplicative scale, obtained by exponentiating the regression coefficient in the log-transformed PCB model. $^k$Adjusted for base variables (maternal age, birthplace, smoking during pregnancy, previous lactation, household income, child’s date of birth and sex, and dredging period). $^l$Represents number of individuals in each food consumption group from a total sample of 531 with nonmissing diet data. $^m$Reflects percent change in a multiplicative scale, obtained by exponentiating the regression coefficient in the log-transformed PCB model. $^n$Adjusted for base variables (maternal age, birthplace, smoking during pregnancy, previous lactation, household income, child’s date of birth and sex, and dredging period). $^o$Represents number of individuals in each food consumption group from a total sample of 531 with nonmissing diet data. $^p$Reflects percent change in a multiplicative scale, obtained by exponentiating the regression coefficient in the log-transformed PCB model. $^q$Adjusted for base variables (maternal age, birthplace, smoking during pregnancy, previous lactation, household income, child’s date of birth and sex, and dredging period). $^r$Represents number of individuals in each food consumption group from a total sample of 531 with nonmissing diet data.
Table 3. Percent change in newborn cord serum PCB levels associated with proxies for maternal dermal and/or inhalation exposure. 

<table>
<thead>
<tr>
<th>Characteristics</th>
<th>ΣPCB</th>
<th>Heavy PCBs</th>
<th>Light PCBs</th>
<th>PCB-118</th>
</tr>
</thead>
<tbody>
<tr>
<td>Occupation</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>None</td>
<td>407</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>≤ 10 years</td>
<td>102</td>
<td>–4</td>
<td>–3</td>
<td>–2</td>
</tr>
<tr>
<td>&gt; 10 years</td>
<td>33</td>
<td>17</td>
<td>16</td>
<td>18</td>
</tr>
<tr>
<td>Gardening</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>None</td>
<td>505</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>≤ 10 years</td>
<td>27</td>
<td>–3</td>
<td>–5</td>
<td>2</td>
</tr>
<tr>
<td>&gt; 10 years</td>
<td>9</td>
<td>–26</td>
<td>–34**</td>
<td>–14</td>
</tr>
<tr>
<td>Other activities</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>None</td>
<td>458</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>≤ 10 years</td>
<td>58</td>
<td>6</td>
<td>7</td>
<td>5</td>
</tr>
<tr>
<td>&gt; 10 years</td>
<td>25</td>
<td>–10</td>
<td>–9</td>
<td>–13</td>
</tr>
<tr>
<td>Residence distance from hot spot (miles)</td>
<td>1.6–2.8</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>0.2–1.5</td>
<td>134</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>1.6–2.8</td>
<td>140</td>
<td>2</td>
<td>2</td>
<td>7</td>
</tr>
<tr>
<td>2.9–3.8</td>
<td>133</td>
<td>4</td>
<td>4</td>
<td>4</td>
</tr>
<tr>
<td>3.9–10.8</td>
<td>135</td>
<td>7</td>
<td>5</td>
<td>15*</td>
</tr>
<tr>
<td>Neighborhood houses built between 1940–1979 (%)</td>
<td>9–20</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>0–20</td>
<td>123</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>21–32</td>
<td>143</td>
<td>2</td>
<td>2</td>
<td>0</td>
</tr>
<tr>
<td>33–60</td>
<td>158</td>
<td>–1</td>
<td>0</td>
<td>–5</td>
</tr>
<tr>
<td>61–100</td>
<td>118</td>
<td>2</td>
<td>2</td>
<td>0</td>
</tr>
</tbody>
</table>

*Reflects percent change in a multiplicative scale, obtained by exponentiating the regression coefficient in the log-transformed PCB model. 

Table 4. Percent change in newborn cord serum PCB levels associated with quartiles of household neighborhood characteristics. 

<table>
<thead>
<tr>
<th>Characteristics</th>
<th>ΣPCB</th>
<th>Heavy PCBs</th>
<th>Light PCBs</th>
<th>PCB-118</th>
</tr>
</thead>
<tbody>
<tr>
<td>Crowding (%)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>0</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>&gt; 0–2</td>
<td>–9</td>
<td>–9</td>
<td>–7</td>
<td>–8</td>
</tr>
<tr>
<td>3–4</td>
<td>–2</td>
<td>–2</td>
<td>–3</td>
<td>–9</td>
</tr>
<tr>
<td>5–16</td>
<td>1</td>
<td>–1</td>
<td>10*</td>
<td>1</td>
</tr>
<tr>
<td>p-Trend</td>
<td>p = 0.78</td>
<td>p = 0.94</td>
<td>p = 0.19</td>
<td>p = 0.70</td>
</tr>
<tr>
<td>Poverty (%)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>0–5</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>6–11</td>
<td>0</td>
<td>1</td>
<td>–6</td>
<td>3</td>
</tr>
<tr>
<td>12–25</td>
<td>–1</td>
<td>–3</td>
<td>1</td>
<td>–1</td>
</tr>
<tr>
<td>26–55</td>
<td>4</td>
<td>–1</td>
<td>15</td>
<td>6</td>
</tr>
<tr>
<td>p-Trend</td>
<td>p = 0.69</td>
<td>p = 0.95</td>
<td>p = 0.11</td>
<td>p = 0.37</td>
</tr>
<tr>
<td>Low income (%)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>0–32</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>33–42</td>
<td>–8</td>
<td>–8</td>
<td>–12</td>
<td>–9</td>
</tr>
<tr>
<td>43–54</td>
<td>7</td>
<td>5</td>
<td>9</td>
<td>2</td>
</tr>
<tr>
<td>55–89</td>
<td>3</td>
<td>0</td>
<td>6</td>
<td>2</td>
</tr>
<tr>
<td>p-Trend</td>
<td>p = 0.32</td>
<td>p = 0.54</td>
<td>p = 0.13</td>
<td>p = 0.20</td>
</tr>
<tr>
<td>Median household income ($)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>7,000–17,000</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>17,001–25,000</td>
<td>12</td>
<td>14</td>
<td>7</td>
<td>–2</td>
</tr>
<tr>
<td>25,001–30,000</td>
<td>–4</td>
<td>–2</td>
<td>–12</td>
<td>–9</td>
</tr>
<tr>
<td>30,001–68,000</td>
<td>–1</td>
<td>3</td>
<td>7</td>
<td>–10%</td>
</tr>
<tr>
<td>p-Trend</td>
<td>p = 0.57</td>
<td>p = 0.90</td>
<td>p = 0.24</td>
<td>p = 0.37</td>
</tr>
<tr>
<td>Low education—less than high school (%)</td>
<td>34–48</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>0–33</td>
<td>–4</td>
<td>–5</td>
<td>–4</td>
<td>1</td>
</tr>
<tr>
<td>49–65</td>
<td>4</td>
<td>1</td>
<td>11</td>
<td>10</td>
</tr>
<tr>
<td>66–80</td>
<td>–6</td>
<td>–9</td>
<td>2</td>
<td>–1</td>
</tr>
<tr>
<td>p-Trend</td>
<td>p = 0.71</td>
<td>p = 0.45</td>
<td>p = 0.42</td>
<td>p = 0.79</td>
</tr>
<tr>
<td>College education (%)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>0–4</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
<td>Reference</td>
</tr>
<tr>
<td>5–9</td>
<td>–4</td>
<td>–1</td>
<td>–11</td>
<td>–1</td>
</tr>
<tr>
<td>10–13</td>
<td>–3</td>
<td>0</td>
<td>–10</td>
<td>0</td>
</tr>
<tr>
<td>14–47</td>
<td>–9*</td>
<td>–5</td>
<td>–17**</td>
<td>–7%</td>
</tr>
<tr>
<td>p-Trend</td>
<td>p = 0.70</td>
<td>p = 0.95</td>
<td>p = 0.37</td>
<td>p = 0.97</td>
</tr>
</tbody>
</table>

*Reflects percent change in a multiplicative scale, obtained by exponentiating the regression coefficient in the log-transformed PCB model. 

Discussion

We found no evidence that living closer to the New Bedford Harbor Superfund site was associated with increased cord serum PCB levels either in the crude unadjusted means or after adjusting for other risk factors for PCB exposure in the study population.

However, children born before or during dredging had consistently higher cord serum PCB levels than children born after dredging, even after we accounted for birth date (Table 5, Figure 2), suggesting a possible effect of the PCB-contaminated site and its dredging on cord blood PCB levels. Serum levels of light PCBs were more strongly associated with dredging than were heavy PCBs (Table 5). This finding suggests that differences in PCB volatility affect exposure risks potentially associated with the site. Furthermore, for PCB-118, a dioxin-like pentachlorinated biphenyl disproportionately prevalent in study samples, the dredging effect was more significant than the temporal decline, with near-constant concentrations before dredging, an increase during dredging, and a significant decline after dredging (Figure 2). Overall, these results support modest, transient increases in cord serum PCB levels during dredging, with significant declines in serum PCB levels observed after dredging, particularly for the more volatile PCBs and PCB-118 (Table 5, Figure 2). The apparent differential effects of remediation on cord serum levels of various congeners are notable given possible congener-specific differences in toxicity.

In addition to the previously described dredging associations, maternal consumption of locally produced dairy products—an exposure risk factor potentially related to the contaminated site—was associated with higher cord serum PCB levels (Tables 2, 5).

The most important predictor of elevated cord serum PCB levels was older maternal age at the birth of the study infant. Older age is a well-established risk factor for increased serum organochlorine concentrations, presumably as a consequence of cumulative exposure and temporal trends in exposure (Kutz et al. 1991). In multivariate models, we found that older maternal age and earlier birth year were both associated with elevated cord blood PCB levels, indicating both cumulative exposure and temporal trend effects.

Mothers born in Portugal, the Azores, or Cape Verde had infants with substantially higher cord serum PCB levels than mothers born in the United States, Canada, or other countries, even after adjustment for diet or other lifestyle covariates that may vary by country of origin (Table 5). Although this...
observed association may be a chance finding or consequent to residual confounding by diet or lifestyle, it is also consistent with potentially higher early-life exposure to PCBs resulting in higher serum levels during pregnancy. For example, PCB contamination is present in fish species from southern Europe and the Atlantic Ocean along the Azores Islands (Stefanelli et al. 2004). Higher early-life exposure to PCBs has been associated with higher serum levels in adulthood among other populations (Rylander et al. 1997).

We confirmed the previously reported association of prior lactation with lower serum PCB levels, which is likely due to PCB excretion in milk (Fitzgerald et al. 1998; Jensen 1991). Smoking during pregnancy was also associated with lower cord serum PCBs. Previous studies have been inconsistent regarding the association of PCBs with smoking; maternal smoking during pregnancy was associated with higher newborn PCB levels in one study (Lackman et al. 2000) but not in others (Fein et al. 1984; Rogan et al. 1986). Smoking may decrease organochlorine concentrations by enhancing their metabolism via smoking-related induction of cytochrome P450 enzymes (Deutch and Hansen 1999; Zevin and Benowitz 1999).

Long-term gardening was associated with lower cord serum PCBs (Table 3), opposite to the hypothesized effect. However, the small number of long-term gardeners (n = 9) suggests that chance and/or confounding may explain this finding. Of note, the final multivariate model (Table 5) was unchanged by the addition of gardening (data not shown).

In addition to these correlates of exposure, maternal consumption of organ meat and local dairy products (Tables 2, 5) was associated with significantly higher cord blood PCB levels, but other potential dietary risks (including fish intake) were not. Although fish and animal products have been identified as important sources of general population exposure to PCBs and dioxins in some studies (Laden et al. 1999; Patandin et al. 1999), levels of PCBs and dioxins in fish and other foods have been declining (Hays and Aylward 2003). Contaminated areas of the harbor were closed to fishing in 1979 (Agency for Toxic Substances and Disease Registry 1995), 14 years before we started this study. A lack of association of local fish consumption with serum PCB levels is consistent with the lag between last likely intake of the most contaminated fish and our exposure assessment.

Other evaluated risk factors did not explain heterogeneity in cord serum PCB levels. Although serum concentrations of PCB-exposed workers are higher than those of the general population (Wolff et al. 1982), the small number of mothers with potential occupational exposure limited the statistical power to detect such associations. Furthermore, the age of study mothers was such that most of their occupational (and other) activities occurred after the ban on PCBs.

Neighborhood socioeconomic status and age of housing were not associated with increased cord PCB levels. Although manufacturers incorporated PCBs in building materials and light fixtures during a well-defined time period (Balfanz et al. 1993), house age was not a good predictor of indoor air PCB concentrations in previous studies in New Bedford (Vorhees et al. 1997). This measure does not capture renovations or other potential indoor PCB sources such as electrical appliances or fluorescent lights. Moreover, the neighborhood distribution of home ages is an imperfect proxy for the age of the specific home of interest.

Correlates of cord serum PCBs did not vary much by the different congener groupings assessed. For example, the exposure pathways we observed for the heavy PCBs and ΣPCBs were quite similar (Table 5). This is likely because the correlation of ΣPCB levels with heavy PCBs was much higher (r = 0.99) than with light PCBs (r = 0.76), consistent with the predominance of heavy PCBs in the sum. Except for diet, correlates of PCB TEQ exposure were also similar to other PCB concentration measures. Specifically, maternal consumption of red meat, but not organ meat, was associated with significantly higher PCB TEQs. Patandin et al. (1999) also found meat to be a major contributor to dietary intake of PCB TEQs. Because the congeners are weighted by dioxin-like activity, these findings provide insights into the correlates of potential toxicity, about which very little is known.

There are several limitations in the interpretation of our findings. First, the median serum PCB level in our cohort was about one-quarter of the overall median in a recent review of 10 studies of PCBs and neurodevelopment (Longnecker et al. 2003). Despite this limitation, our findings corroborate previously established correlates of serum PCB levels, including age and secular trends. In addition, the use of simplified proxies for some exposure pathways limited our ability to determine the relative contribution of various routes of exposure to cord serum PCB levels. In particular, it could be argued that residential distance from the site does not capture outdoor concentrations because it ignores prevailing winds. However, the maps of cord serum PCB levels and model residuals do not indicate any likely wind-related spatial patterns with this region’s prevailing wind direction from the south-southwest (Cullen et al. 1996). In addition, it is possible that including household income and other demographic variables reduced our ability to characterize exposure pathways by overadjusting for these indirect correlates of exposure. However, sensitivity analyses demonstrated that this was not the case. For example, results of our pathway analyses were not substantially changed by removing income from the model. Lastly, the cross-sectional

Table 5. Percent changea (95% confidence interval) in newborn cord serum PCB levels as a function of significant maternal and infant predictors (p < 0.10).

<table>
<thead>
<tr>
<th>Child characteristics</th>
<th>ΣPCB (n = 541, R2 = 38%)</th>
<th>Heavy PCBs (n = 541, R2 = 37%)</th>
<th>Light PCBs (n = 541, R2 = 28%)</th>
<th>PCB-118 (n = 541, R2 = 30%)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Male sex</td>
<td>–7 (–15 to 3)</td>
<td>–7 (–16 to 2)</td>
<td>–1 (–11 to 10)</td>
<td>–8 (–19 to 4)</td>
</tr>
<tr>
<td>Date of birth (years)b</td>
<td>–36 (–52 to –13)**</td>
<td>–31 (–50 to –6)**</td>
<td>–43 (–50 to –21)**</td>
<td>–13 (–41 to 27)**</td>
</tr>
<tr>
<td>Child born before/during drugging</td>
<td>17 (3–3 to 40)**</td>
<td>16 (–4 to 41)**</td>
<td>26 (2 to 59)**</td>
<td>39 (10 to 75)**</td>
</tr>
<tr>
<td>Mother’s demographics</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Maternal age (years)b</td>
<td>36 (29 to 43)**</td>
<td>40 (33 to 47)**</td>
<td>17 (11 to 24)**</td>
<td>25 (17 to 33)**</td>
</tr>
<tr>
<td>Born outside United States/Canada</td>
<td>42 (23 to 63)**</td>
<td>41 (21 to 63)**</td>
<td>42 (21 to 67)**</td>
<td>69 (42 to 102)**</td>
</tr>
<tr>
<td>Portugal/Azores/Cape Verde</td>
<td>20 (–4 to 48)**</td>
<td>20 (–4 to 50)**</td>
<td>20 (–6 to 54)**</td>
<td>33 (1 to 74)**</td>
</tr>
<tr>
<td>Previous lactation (&gt;6months)</td>
<td>–25 (–35 to –12)**</td>
<td>–25 (–36 to –12)**</td>
<td>–26 (–38 to –12)**</td>
<td>–33 (–45 to 19)**</td>
</tr>
<tr>
<td>Smoking during pregnancy (yes)</td>
<td>–11 (–20 to –1)**</td>
<td>–12 (–21 to 2)**</td>
<td>–11 (–21 to 1)**</td>
<td>–29 (–38 to 19)**</td>
</tr>
<tr>
<td>Household income ($40000)</td>
<td>1 (–13 to 10)</td>
<td>–4 (–15 to 9)</td>
<td>5 (–6 to 20)</td>
<td>4 (–10 to 21)</td>
</tr>
<tr>
<td>Mother’s diet</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Organ meat (&gt;1/month)</td>
<td>21 (2 to 44)**</td>
<td>19 (–1 to 42)**</td>
<td>30 (7 to 59)**</td>
<td>30 (5 to 62)**</td>
</tr>
<tr>
<td>Local dairy (yes)</td>
<td>19 (1 to 39)**</td>
<td>19 (1 to 41)**</td>
<td>8 (–10 to 29)</td>
<td>16 (–6 to 42)</td>
</tr>
<tr>
<td>Red meat (&gt;2/week)</td>
<td>7 (–5 to 19)</td>
<td>5 (–6 to 18)</td>
<td>12 (–1 to 27)**</td>
<td>12 (–2 to 29)</td>
</tr>
<tr>
<td>Dark fish (&gt;1/month)</td>
<td>8 (–4 to 21)</td>
<td>7 (–6 to 21)</td>
<td>13 (–3 to 30)**</td>
<td>14 (–2 to 33)**</td>
</tr>
</tbody>
</table>

*aReflects percent change in a multiplicative scale, obtained by exponentiating the regression coefficient in the log-transformed PCB model. **p < 0.05; *p < 0.01.
nature of this analysis limits the certainty of inferences regarding the observed temporal- and dredging-associated differences in serum PCB levels.

In conclusion, our findings among New Bedford area infants suggest that maternal residence near a Superfund site per se does not lead to higher cord serum PCB levels independent of other exposure risk factors, such as maternal age, birthplace, diet, previous lactation, pregnancy smoking, and infant date of birth. However, there was evidence of an important local impact on exposure risk as shown by increased cord serum PCB levels in association with maternal local dairy consumption and lower cord serum PCB levels after site dredging.

References


