

1 **Exposure to elemental composition of outdoor PM_{2.5} at birth and cognitive and**
2 **psychomotor function in childhood in four European birth cohorts**

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52

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93 **Abstract**

94 **Background:** Little is known about developmental neurotoxicity of particulate matter
95 composition. We aimed to investigate associations between exposure to elemental
96 composition of outdoor PM_{2.5} at birth and cognitive and psychomotor functions in childhood.

97 **Methods:** We analyzed data from 4 European population-based birth cohorts in the
98 Netherlands, Germany, Italy and Spain, with recruitment in 2000-2006. Elemental
99 composition of PM_{2.5} measurements were performed in each region in 2008-2011 and land
100 use regression models were used to predict concentrations at participants' residential
101 addresses at birth. We selected 8 elements (copper, iron, potassium, nickel, sulfur, silicon,
102 vanadium and zinc) and used principal component analysis to combine elements from the
103 same sources. Cognitive (general, verbal, and non-verbal) and psychomotor (fine and gross)
104 functions were assessed between 1 and 9 years of age. Adjusted cohort-specific effect
105 estimates were combined using random-effects meta-analysis.

106 **Results:** 7,246 children were included in this analysis. Single element analysis resulted in
107 negative association between estimated airborne iron and fine motor function (-1.25 points
108 [95% CI -2.45 to -0.06] per 100 ng/m³ increase of iron). Association between the motorized
109 traffic component, derived from principal component analysis, and fine motor function was
110 not significant (-0.29 points [95% CI -0.64 to 0.06] per unit increase). None of the elements
111 were associated with gross motor function or cognitive function, although the latter estimates
112 were predominantly negative.

113 **Conclusion:** Our results suggest that iron, a highly prevalent element in motorized traffic
114 pollution, may be a neurotoxic compound. This raises concern given the ubiquity of
115 motorized traffic air pollution.

116 **Keywords:** air pollution; particulate matter; neuropsychological tests; child health; child
117 development

118 **1. Introduction**

119 Air pollution is a serious threat to human health. The potential effects of air pollution on
120 human brain are an active area of research (Block et al., 2012). Particulate matter (PM),
121 highly prevalent in traffic related air pollution, could reach the brain and other organs by
122 translocation to the systemic circulation following a deposition in the pulmonary region after
123 inhalation (Block et al., 2012). The brain of a fetus could be reached via an indirect path as
124 the placenta and the blood-brain barrier grant only a partial protection against entry of
125 environmental toxicants to which the mother is exposed. As the brain is in the process of
126 development and the detoxification mechanisms are relatively immature, the potential adverse
127 effects of exposure to air pollution during pregnancy are of particular concern (Block et al.,
128 2012; Grandjean and Landrigan, 2014).

129 Although the precise biological mechanisms are yet to be clarified, there is some
130 evidence for a negative association between pre- and postnatal exposure to outdoor PM and
131 children’s cognition, psychomotor development, and behavioral problems (Guxens and
132 Sunyer, 2012; Guxens et al., 2014, 2015; Suades-González et al., 2015). It has been
133 hypothesized that traffic-related PM might be neurotoxic mainly through some of its
134 components such as polycyclic aromatic hydrocarbons (PAHs), black carbon, and trace
135 elements, potentially leading to increased oxidative stress and increased activation of brain
136 microglia, the primary regulators of neuroinflammation (Block et al., 2012). Studies focusing
137 on PAHs found negative association with children’s cognition and behavioral problems
138 (Edwards et al., 2010; Lovasi et al., 2014; Perera et al., 2006, 2009, 2013; Wang et al., 2010).
139 Moreover, a recent study using magnetic resonance imaging found preliminary evidence for
140 reduction in the white matter surface of the left hemisphere of the brain in childhood with
141 increased prenatal concentrations of PAHs, associated with slower information processing
142 speed (Peterson et al., 2015). Studies with focus on pre- and postnatal exposure to black

143 carbon also found a negative association with cognitive and/or psychomotor development
144 (Chiu et al., 2013; Suglia et al., 2008), although these findings were inconsistent.

145 To date, developmental neurotoxicity has been documented for only a small selection
146 of existing trace elements (Grandjean and Landrigan, 2014). Studies addressing the
147 association between pre- and/or postnatal exposure to trace elements in outdoor air and
148 children's brain development are very limited in number. The few existing studies have linked
149 higher levels of several airborne elements including arsenic, cadmium, chromium, lead,
150 manganese, mercury, nickel, selenium, and vanadium, to elevated prevalence of autism
151 spectrum disorder (Lam et al., 2016). Additionally, the only study to date that focused on
152 airborne elements and cognition, found evidence for a negative association between childhood
153 exposure at schools to airborne elements originating from motorized traffic sources and
154 specific cognitive functions in school aged children (Basagaña et al., 2016). However, for
155 many elements, sparse evidence of neurotoxicity is possibly a consequence of limited amount
156 of research addressing the topic rather than absence of an association (Grandjean and Herz,
157 2015).

158 Therefore, the aim of this study was to analyze the association between exposure at
159 birth to a set of elements measured in outdoor PM with aerodynamic diameter of less than 2.5
160 micrometers (PM_{2.5}) and cognitive and psychomotor function in childhood using data from
161 four European cohorts. The elemental components examined in this study were copper, iron,
162 potassium, nickel, sulfur, silicon, vanadium and zinc, selected based on their reflection of
163 major anthropogenic emission sources. This study builds on a previous epidemiological study
164 that investigated the association between air pollution and neuropsychological development in
165 6 European cohorts (Guxens et al., 2014). In that study, the authors found a negative
166 association between prenatal exposure to NO₂ and PM - latter borderline significant - and
167 psychomotor function in childhood. The cohorts included in the current study are a subset of

168 the cohorts studied previously due to the availability of elemental composition data. Also, in
169 the current study we used additional neuropsychological domains and some of the tests
170 included were carried out at older ages.

171

172 **2. Methods**

173 ***2.1. Population and Study Design***

174 This study is part of the ESCAPE (European Study of Cohorts for Air Pollution Effects;
175 www.escapeproject.eu) project. The aim of the project was to investigate the association
176 between exposure to outdoor air pollution and health within prospective cohort studies. In the
177 current study, we included 4 European population-based birth cohorts: GENERATION R
178 (The Netherlands) (Jaddoe et al., 2012), DUISBURG (Germany) (Wilhelm et al., 2008),
179 GASPII (Italy) (Kooijman et al., 2016), and INMA-Sabadell (Spain) (Guxens et al., 2012), a
180 selection based on the availability of elemental composition of PM_{2.5} and neuropsychological
181 data. Mother-child pairs were recruited between 2000 and 2006. A total of 7,246 children
182 aged between 1 and 9 years was included in this analysis and had data on exposures and at
183 least one of the neuropsychological outcomes (Table 1). Local authorized Institutional
184 Review Boards granted the ethical approval for the studies and all participants provided
185 signed informed consent.

Table 1. Description of the Participating Birth Cohort Studies

Origin of the study (city/area)	Setting			Elemental Components	Cognitive function					Psychomotor function				
	Cohort name	Birth period	No. of participants at baseline	LUR models available	Test	Domain	Age (years)	Evaluator	No. ^a	Test	Domain	Age (years)	Evaluator	No. ^a
Dutch (Rotterdam)	Generation R	2001-2006	8737	Cu, Fe, K, Ni, S, Si, V, Zn	MCDI	verbal	1.5	Parents	4397	MIDI	FM, GrM	1	Parents	4704
					SON-R	non-verbal	6	Trained staff	4580					
German (Ruhr area)	Duisburg	2000-2004	232	Cu, Fe, Ni, S, Si, V, Zn	BSID II	general cognition	1	Psychologist	186	N/A				
					BSID II	general cognition	2	Psychologist	178					
					HAWIK-IV	general cognition, verbal, non-verbal	9	Psychologist	95					
Italian (Rome)	GASPII	2003-2005	719	Cu, Fe, K, Ni, S, Si, V, Zn	DDST II	verbal	1.5	Parents	546	DDST II	FM, GrM	1.5	Pediatrician	546
					WISC-III	general cognition, verbal, non-verbal	7	Psychologist	450					
Spanish (Sabadell)	INMA-Sabadell	2004-2007	740	Cu, Fe, K, Ni, S, Si, V, Zn	BSID I	general cognition	1.5	Psychologist	519	MSCA	FM, GrM	4	Psychologist	439
					MSCA	general cognition, verbal, non-verbal	4	Psychologist	439					

BSID, Bayley Scales of Infant Development (I-first edition, II-second edition); DDST II, Denver Developmental Screening Test II; FM, Fine motor; GrM, Gross motor; HAWIK-IV, Hamburg Wechsler Intelligenztest für Kinder - IV; MCDI, McArthur Communicative Development Inventory; MIDI, Minnesota Infant Development Inventory; MSCA, McCarthy Scales of Children's Abilities; N/A, not available; SON-R, De Snijders-Oomen Niet-verbale Intelligenztest-Revisie; WISC, Wechsler Intelligence Scale for Children

^aNumber of subjects with airborne elemental components of PM_{2.5} and cognitive/psychomotor function data available

187 *2.2. Exposure to Elemental Composition of Outdoor PM_{2.5} at birth*

188 The exposure of each participant to the elemental composition of PM_{2.5} at birth was estimated
189 using standardized procedure based on land use regression (LUR) methodology (de Hoogh et
190 al., 2013). A number of studies have shown that LUR models provide a cost-effective
191 methodology to capture the spatial contrasts of air pollution (Hoek et al., 2008; Marshall et
192 al., 2008). The locations of the measuring stations were based on the specific characteristics
193 of each study area including a large diversity of potential sources of air pollution variability,
194 and were selected in a manner to maximise the representativeness of the residential addresses
195 of the cohort participants (Eeftens et al., 2012). We focused on fine particles rather than
196 coarse, due to their higher potential to translocate to the systemic circulation because of the
197 smaller size (Phalen et al. 2010). PM_{2.5} concentrations in outdoor air were measured at 40
198 sites in the Netherlands/Belgium and Catalunya, and 20 sites in Ruhr area and Rome three
199 times over a year (in summer, winter, and an intermediate season) during a two-week period
200 each time to capture seasonal variations (Eeftens et al., 2012). The campaigns took place
201 between 2008 and 2011. The filters were sent to Cooper Environmental Services (Portland,
202 Oregon, USA) to analyze their elemental composition using X-Ray Fluorescence (XRF) (de
203 Hoogh et al., 2013; Tsai et al., 2015). The results of the three measurements were then
204 averaged, adjusting for temporal trends using data from a continuous reference site, resulting
205 in one mean annual concentration for each element identified in the composition of PM_{2.5}.

206 Following previous ESCAPE studies on elemental components (de Hoogh et al., 2013;
207 Pedersen et al., 2016; Wang et al., 2014) we selected 8 elements based on their reflection of
208 major anthropogenic emission sources and on data availability determined by (i) the
209 coefficient of variation acquired from duplicate samples, (ii) the percentage of samples in
210 which the element was detected, and (iii) the availability of relevant geographical data needed
211 as predictor variables in the LUR models. Copper, iron, and zinc reflect brake linings, tire

212 wear, and industrial (smelter) emissions, silicon and potassium reflect crustal materials and
213 biomass burning, and fossil fuel combustion is reflected by nickel, vanadium, and sulfur
214 (Viana et al., 2008).

215 Following a previous study on birth outcomes (Pedersen et al., 2016), concentrations
216 of the selected elements were assigned at each participants' home address at birth to obtain an
217 estimation of the pregnancy exposure using mean annual area-specific LUR model estimates
218 based on 2008-2011 data (Table 1). Fixed increments per elemental component were applied
219 to facilitate comparability. The model predictors and a description of model performances are
220 reported elsewhere (de Hoogh et al., 2013). Due to insufficient data quality, LUR models of
221 potassium could not be developed for the German cohort (Table 1). Next, we pooled the
222 exposure data of participants from the cohorts together and applied principal component
223 analysis (PCA) to the estimated elemental concentrations at the residential addresses, in order
224 to combine elements from the same sources into one score. Oblique promax rotations were
225 allowed. Since the levels of potassium could not be estimated for the German cohort, that
226 cohort was not included in the pooled PCA.

227

228 ***2.3. Cognitive and Psychomotor Function***

229 Neuropsychological tests used to assess the cognitive and psychomotor function of children
230 were administered by psychologists, pediatricians or trained research staff, or by
231 questionnaires answered by the parents, and differed between the cohorts (Table 1). For each
232 cohort, the tests and questionnaires that measured each neuropsychological function in a
233 similar way and derived in comparable score distribution, were selected. Cognitive function
234 scales measured general, verbal, and/or non-verbal cognitive functions and psychomotor
235 function scales measured fine and gross motor functions (Table 1). To homogenize the scales,
236 we converted all raw scores into standard deviation units using the z -score (z -score is

237 calculated as the raw score minus the sample mean, divided by the standard deviation) and
238 standardized them to a mean of 100 and a standard deviation of 15 (new score = 100 + (15 ×
239 z)) (Guxens et al., 2014). For each domain, higher scores corresponded to better
240 neuropsychological function.

241

242 ***2.4. Potential confounding variables***

243 Available potential confounding variables were defined a priori based on direct acyclic graph
244 (DAG) (Supplementary Material Figure 1) and selected as similarly as possible across the
245 cohorts. Maternal information included age at delivery (continuous in years), height
246 (continuous in centimeters), pre-pregnancy body mass index (continuous in kg/m²), smoking
247 during pregnancy (yes or no), alcohol consumption during pregnancy (yes or no), marital
248 status (monoparental household: yes or no), and parity (0, 1, ≥2). Parental information
249 included educational level (low, medium, high) and country of birth (country of the cohort or
250 foreign country). Maternal height and pre-pregnancy weight were obtained at the enrollment
251 in the study, or self-reported in the first trimester of the pregnancy, at birth or two weeks after
252 birth of the child. The other variables were collected through questionnaires either during
253 pregnancy or at birth. For education level, standardization of cohort-specific categories was
254 applied to create a common variable (Guxens et al., 2014). Child's age at the time of the
255 cognitive and psychomotor function assessment, and the evaluator for the assessment, were
256 also recorded.

257

258 ***2.5. Statistical Analyses***

259 We applied multiple imputation of missing values using chained equations to impute missing
260 potential confounding variables among all participants with available data on exposure and at
261 least one outcome variable (Supplementary Material Table 1). We obtained 25 completed

262 datasets that we analyzed using standard procedures for multiple imputation (Spratt et al.,
263 2010; Sterne et al., 2009). Children with available exposure and outcome data (n=7,246) were
264 more likely to have parents with higher socioeconomic status compared to those recruited
265 initially in the cohorts but without available data on exposure and outcome (n=3,182)
266 (Supplementary Material Tables 2 and 3). We used inverse probability weighting (IPW) to
267 correct for loss to follow-up, i.e. to account for selection bias that potentially arises when only
268 population with available exposure and outcome data, and here thus with relatively higher
269 socioeconomic status, is included as compared to a full initial cohort recruited at pregnancy
270 (Weisskopf et al., 2015; Weuve et al., 2012). Briefly, we used information available for all
271 participants at recruitment to predict the probability of participation in the study, and used the
272 inverse of those probabilities as weights in the analyses so that results would be representative
273 for the initial populations of the cohorts. The variables used to create the weights are
274 described in Supplementary Material Table 4.

275 After visual inspection for linearity, we used linear regression models to analyze the
276 relationships of each single element and PCA component with each neuropsychological
277 function. Additionally, we performed the analyses with prenatal PM_{2.5} and NO₂ levels and
278 each neuropsychological function to make the comparison with the previous study (Guxens et
279 al., 2014) straightforward. Concentrations of the pollutants were introduced as continuous
280 variables and were not transformed. When the age of a child was not linearly related with
281 cognitive or psychomotor function scale, we used the best transformation of age found using
282 fractional polynomials (Royston et al., 1999). The models were adjusted for all potential
283 confounding variables described in the previous sub-chapter.

284 We carried out a two-steps analysis. First, associations were analyzed separately for
285 each cohort. Second, cohort-specific effect estimates were combined in a meta-analysis.
286 Because the data originated from four different regions with divergent characteristics, we

287 decided to use a conservative approach selecting a priori random effect meta-analysis method
288 thereby also adding to the homogeneity and comparability of the analyses. We used Cochran
289 Q test and I^2 statistic to indicate total variability in the estimates that is attributable to
290 between-cohort heterogeneity (Higgins and Thompson, 2002). When the same outcome was
291 measured at multiple ages in a cohort, the score at the oldest age was taken into account in the
292 meta-analysis. Exception was made for the general cognitive function in the German cohort
293 wherein the second oldest age was selected due to substantially larger sample size compared
294 to the sample size of the oldest age (Table 1). Finally, to test the sensitivity of the results, we
295 repeated the meta-analyses including younger ages among the cohorts where the outcomes
296 were measured at different ages, as well as including the oldest age for the German cohort.
297 All statistical hypothesis tests were two-tailed with significance level set at $p < 0.05$ and were
298 carried out using STATA (version 14.0; StataCorporation, College Station, TX).

299

300 **3. Results**

301 Parental characteristics of the study population are shown in Table 2. The percentage of
302 higher-educated mothers was highest in the Dutch cohort while the percentage of higher-
303 educated fathers was highest in the German cohort. The highest percentage of both - lower
304 educated mothers and lower educated fathers - was in the Spanish cohort. The highest
305 percentage of mothers consuming alcohol during pregnancy was in the Dutch cohort whereas
306 the highest percentage of mothers smoking during pregnancy was in the Spanish cohort. The
307 proportion of parents that were born in a country different than that of the study and the
308 percentage of single parent households was highest in the Dutch cohort.

309 Cohort specific concentration levels of each element are shown in Figure 1.

310 Correlations between the modelled concentrations of the pollutants varied considerably
311 depending on the pollutant and the region (Supplementary Material Table 5). The PCA

312 resulted in identification of two principal components with a combined R^2 of 78% and a low
313 correlation of <0.20 . Component 1 was loaded primarily with copper, iron and sulfur
314 suggesting a reflectance of motorized traffic pollution. Component 2 was comprised
315 predominantly of positive loadings of nickel and vanadium and negative loadings of silicon
316 and potassium, making the conceptualization of component 2 highly ambiguous
317 (Supplementary Material Table 6). Therefore, this component was not analyzed further. The
318 proportion of participants with a higher socioeconomic status was larger in areas with higher
319 levels of motorized traffic pollution expressed in tertiles (Supplementary Material Table 7).

320 None of the elemental components nor the motorized traffic pollution source was
321 associated with cognitive functions (general, verbal, and non-verbal) (Table 3 and
322 Supplementary Material Figures 2-4). Effect estimates were predominantly negative,
323 suggesting that higher pollution levels at birth were associated with a lower cognitive
324 function, but significance was not reached in any of the associations (e.g. lower score by 1.68
325 points in general cognitive function [95% confidence interval (CI): -5.08 to 1.72] for each 5
326 ng/m^3 increase in the prenatal levels of airborne copper). Increase in the levels of airborne
327 iron, one of the main elements of motorized traffic component, was negatively associated
328 with fine motor function (lower score by 1.25 points [95% CI -2.45 to -0.06] for each 100
329 ng/m^3 increase of iron), but the association between motorized traffic component and fine
330 motor function was not significant (-0.29 [95% CI -0.64 to 0.06] for each one unit increase of
331 the component) (Table 4 and Figure 2). We did not find any associations between the
332 elemental components or the motorized traffic pollution component and gross motor function
333 (Table 4 and Supplementary Material Figure 5). Overall higher exposure to NO_2 and $\text{PM}_{2.5}$
334 was related to a decrease in general cognition and fine motor function, but also here no
335 significance was reached in any of the associations (Supplementary Material Table 8 and
336 Figures 6-10). Region-specific effect estimates were relatively homogeneous except for some

337 results obtained in the analyses with nickel and vanadium where evidence for heterogeneity
338 between the cohorts was observed (Table 3 and Table 4). Sensitivity analyses selecting
339 younger ages, and the analyses wherein older ages in the German cohort were included,
340 produced only minimal changes in the results (Supplementary Material Table 9).

Table 2. Distribution of Parental Characteristics

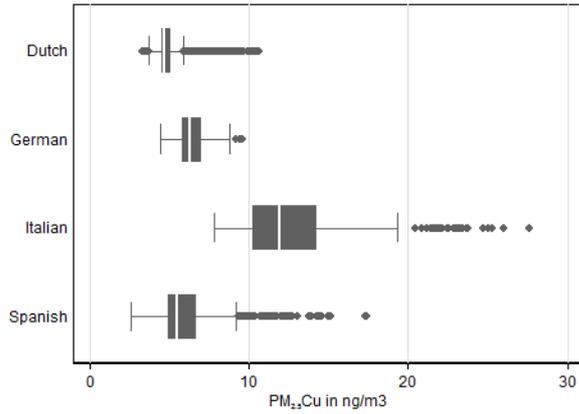
Study				Maternal Education Level			Paternal Education Level			Maternal country of Birth	Paternal country of Birth	Maternal Age at Delivery
Origin	Country	Name	No.	Low	Medium	High	Low	Medium	High	Foreign	Foreign	(in years)
Dutch	The Netherlands	Generation R	5911	9.3	41.9	48.8	9.6	41.8	48.6	44.3	42.1	31.0 (5.0)
German	Germany	Duisburg	190	20.0	37.9	42.1	24.6	24.2	51.3	13.2	18.4	31.2 (4.7)
Italian	Italy	GASPII	614	13.6	50.7	35.8	1.7	66.5	31.7	3.8	2.4	33.4 (4.4)
Spanish	Spain	INMA-Sabadell	531	27.4	41.5	31.1	35.9	42.7	21.4	10.2	11.4	31.7 (4.2)

Study				Maternal Pre-Pregnancy Body Mass Index	Maternal Height	Maternal Alcohol During Pregnancy	Maternal Smoking During Pregnancy	Parity	Marital Status
Origin	Country	Name	No.	(in kg/m²)	(in cm)	Yes	Yes	Nulliparous	Monoparental
Dutch	The Netherlands	Generation R	5911	22.6 (20.8 to 25.2)	168.0 (7.4)	42.5	14.3	57.1	11.2
German	Germany	Duisburg	190	22.9 (20.8 to 25.7)	167.6 (6.2)	11.5	22.6	56.8	2.7
Italian	Italy	GASPII	614	21.3 (19.8 to 23.7)	164.8 (5.8)	35.6	11.3	58.0	0.5
Spanish	Spain	INMA-Sabadell	531	22.7 (21.0 to 25.4)	162.4 (6.0)	21.1	29.4	57.3	1.1

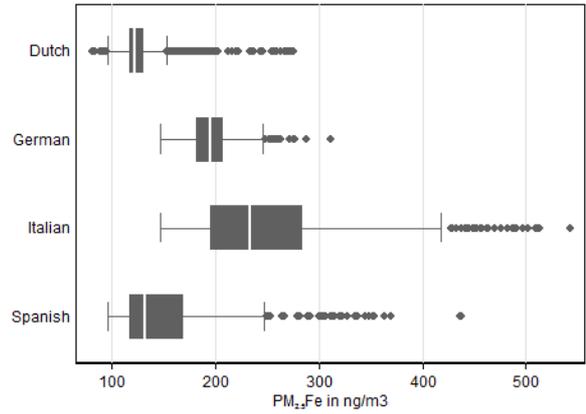
Values are percentages for the categorical variables, mean (standard deviation) for the continuous normally distributed variables, and median (interquartile range) for the non-normally distributed variables.

No=number

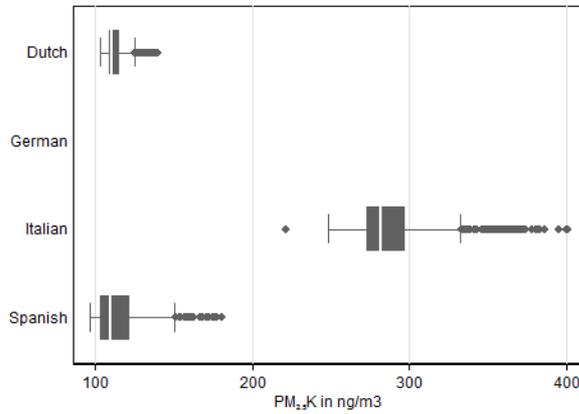
A. Distribution of copper (in ng/m^3)



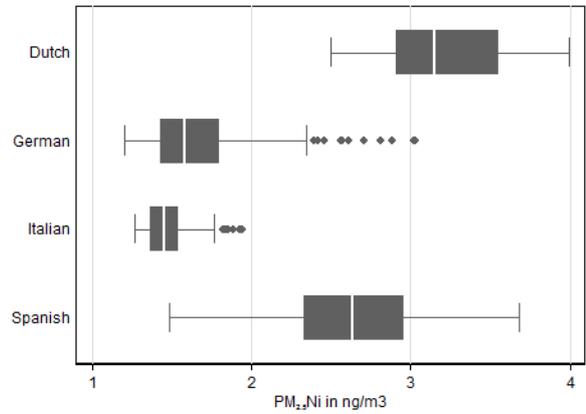
B. Distribution of iron (in ng/m^3)



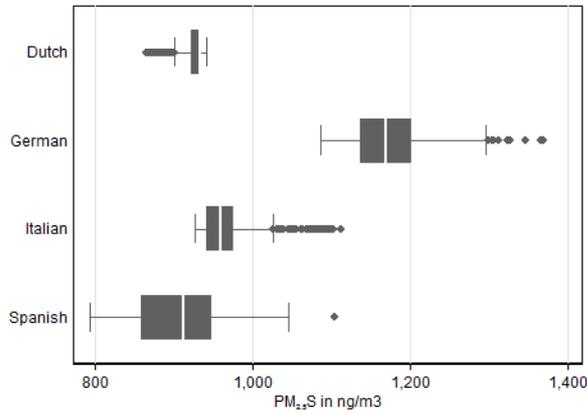
C. Distribution of potassium (in ng/m^3)



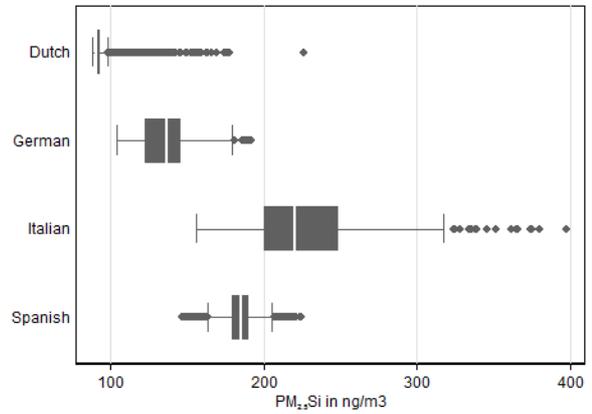
D. Distribution of nickel (in ng/m^3)

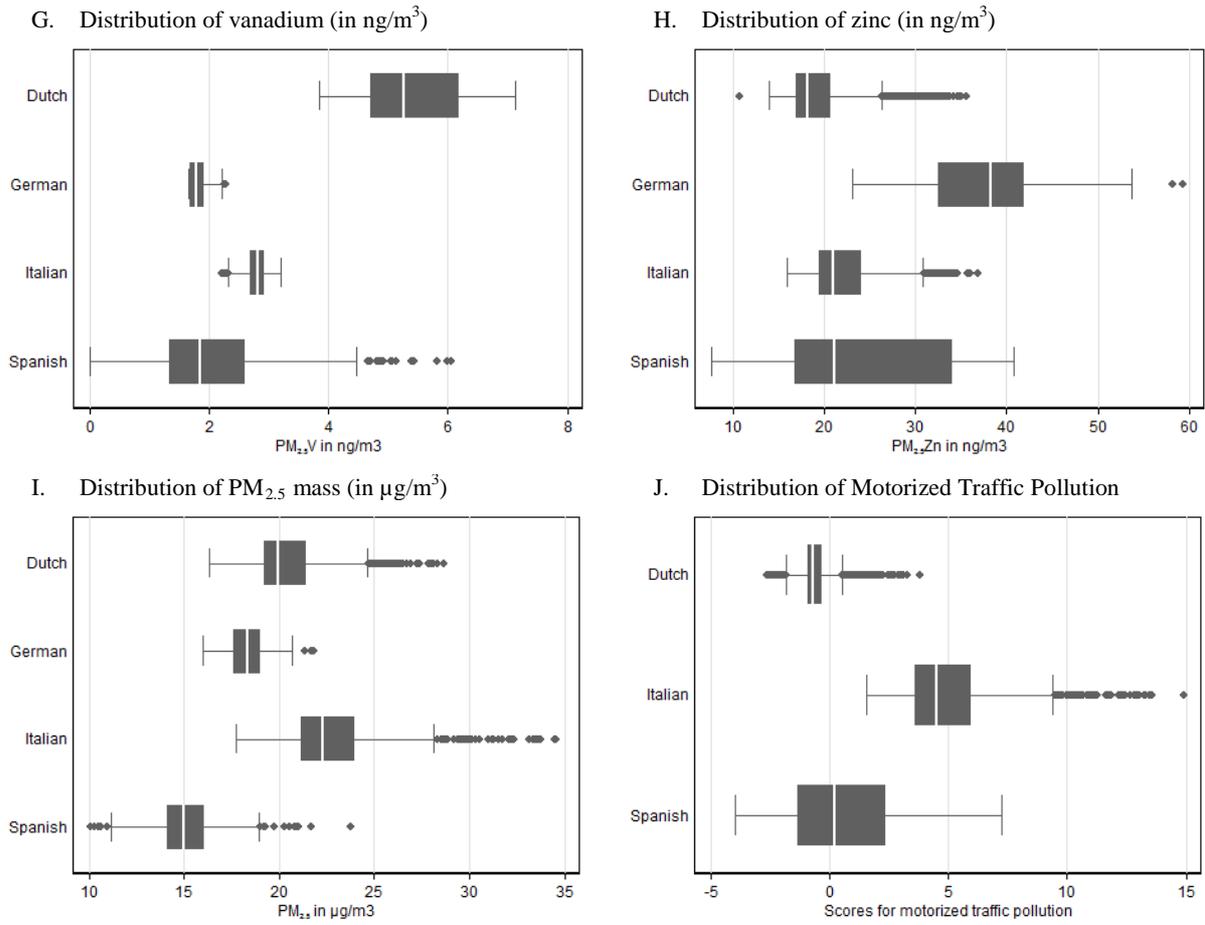


E. Distribution of sulfur (in ng/m^3)



F. Distribution of silicon (in ng/m^3)





342 **Figure 1. Distribution of PM_{2.5} elemental composition levels in ng/m³ (copper (A), iron (B), potassium (C),**
 343 **nickel (D), sulfur (E), silicon (F), vanadium (G) and zinc (H)), PM_{2.5} mass in µg/m³ (I) and motorized**
 344 **traffic pollution scores (J) in the participating cohorts**

Table 3. Fully adjusted combined associations between exposure to elemental components and the identified pollution component at birth, and general, verbal, and non-verbal cognitive function

	General cognitive function			Verbal cognitive function			Non-verbal cognitive function		
	Coef. (95% CI)	Test for Heterogeneity, P	I ² (%)	Coef. (95% CI)	Test for Heterogeneity, P	I ² (%)	Coef. (95% CI)	Test for Heterogeneity, P	I ² (%)
PM_{2.5} Cu	-1.68 (-5.08 to 1.72)	0.066	63.2	-0.27 (-1.58 to 1.04)	0.710	0.0	-1.03 (-2.35 to 0.30)	0.385	1.4
PM_{2.5} Fe	-1.26 (-3.21 to 0.70)	0.201	37.7	-0.28 (-1.47 to 0.91)	0.836	0.0	-1.09 (-2.28 to 0.11)	0.684	0.0
PM_{2.5} K	-0.87 (-2.79 to 1.06)	0.309	3.2	0.01 (-1.62 to 1.63)	0.890	0.0	-0.65 (-2.74 to 1.43)	0.222	33.5
PM_{2.5} Ni	-1.99 (-4.59 to 0.62)	0.784	0.0	-0.22 (-1.31 to 0.87)	0.611	0.0	-0.17 (-1.48 to 1.14)	0.357	7.2
PM_{2.5} S	-2.40 (-5.85 to 1.05)	0.746	0.0	-2.75 (-5.92 to 0.43)	0.722	0.0	-0.48 (-3.71 to 2.76)	0.374	3.7
PM_{2.5} Si	-2.54 (-5.53 to 0.45)	0.791	0.0	-1.19 (-3.36 to 0.98)	0.883	0.0	-0.77 (-2.95 to 1.41)	0.861	0.0
PM_{2.5} V	-4.56 (-12.02 to 2.89)	0.210	35.9	-0.24 (-1.22 to 0.73)	0.419	0.0	-1.04 (-4.55 to 2.47)	0.020	69.6
PM_{2.5} Zn	-0.66 (-1.87 to 0.55)	0.504	0.0	-0.05 (-0.92 to 0.81)	0.918	0.0	0.09 (-0.74 to 0.92)	0.704	0.0
Motorized traffic component^a	-0.26 (-0.65 to 0.14)	0.426	0.0	-0.16 (-0.51 to 0.19)	0.670	0.0	-0.20 (-0.55 to 0.15)	0.677	0.0

Coef=coefficient; CI=confidence intervals; Cu=copper; Fe=iron; I²=percentage of the total variability due to between-areas heterogeneity; K=potassium; Ni=nickel; S=sulfur; Si=silicon; V=vanadium; Zn=zinc.

^aMotorized traffic component was acquired using the principle component analysis (PCA). See Supplementary Table 6 for detailed configuration of the component.

Coefficient and 95% CI were estimated by random-effects meta-analysis by cohort. Models were adjusted for parental education levels, parental countries of origin, maternal age at delivery, maternal pre-pregnancy body mass index, maternal height, maternal alcohol consumption during pregnancy, maternal smoking during pregnancy, marital status, parity and age of the child at neuropsychological testing per increments of 5 ng/m³ for Cu PM_{2.5}; 100 ng/m³ for Fe PM_{2.5}; 50 ng/m³ for K PM_{2.5}; 1 ng/m³ for Ni PM_{2.5}; 200 ng/m³ for S PM_{2.5}; 100 ng/m³ for Si PM_{2.5}; 2 ng/m³ for V PM_{2.5}; 10 ng/m³ for Zn PM_{2.5}; and 1 unit for motorized traffic component

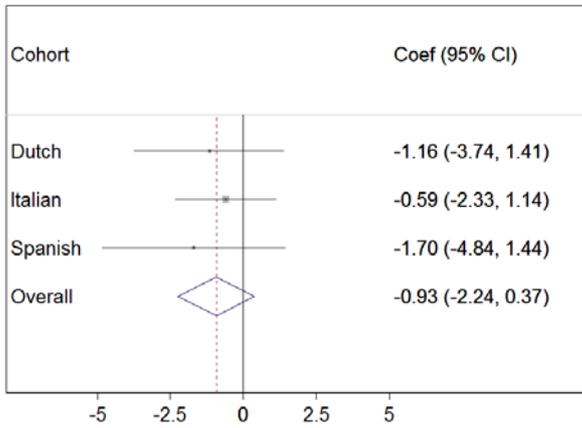
Table 4. Fully adjusted combined associations between exposure to elemental components and the identified pollution source at birth, and fine and gross motor function

	Fine motor function			Gross motor function		
	Coef. (95% CI)	Test for Heterogeneity, P	I ² (%)	Coef. (95% CI)	Test for Heterogeneity, P	I ² (%)
PM_{2.5} Cu	-0.93 (-2.24 to 0.37)	0.816	0.0	0.29 (-2.15 to 2.72)	0.052	66.2
PM_{2.5} Fe	-1.25 (-2.45 to -0.06)	0.784	0.0	-0.08 (-1.93 to 1.77)	0.108	55.1
PM_{2.5} K	-1.10 (-2.69 to 0.48)	0.390	0.0	1.03 (-0.61 to 2.67)	0.353	4.0
PM_{2.5} Ni	-1.11 (-5.91 to 3.69)	0.003	83.1	1.86 (-0.76 to 4.49)	0.147	47.9
PM_{2.5} S	-1.01 (-4.32 to 2.29)	0.598	0.0	1.76 (-1.59 to 5.12)	0.776	0.0
PM_{2.5} Si	-1.52 (-3.61 to 0.57)	0.970	0.0	-1.64 (-3.69 to 0.40)	0.455	0.0
PM_{2.5} V	-0.86 (-4.93 to 3.22)	0.003	82.4	-0.22 (-4.18 to 3.74)	0.007	79.7
PM_{2.5} Zn	-0.36 (-1.22 to 0.51)	0.655	0.0	0.59 (-0.59 to 1.77)	0.217	34.5
Motorized traffic component^a	-0.29 (-0.64 to 0.06)	0.879	0.0	0.10 (-0.42 to 0.62)	0.120	52.8

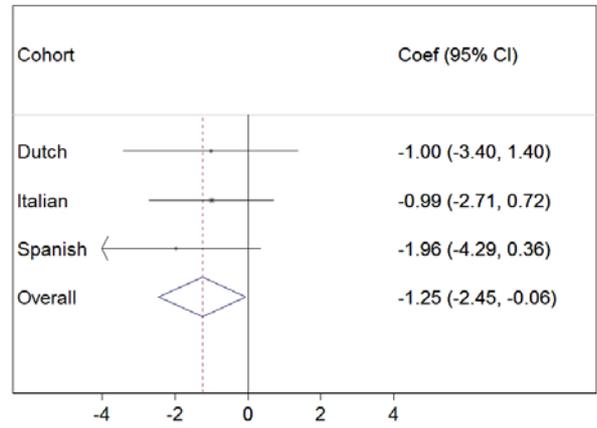
Coef= coefficient; CI=confidence intervals; Cu=copper; Fe=iron; I²=percentage of the total variability due to between-areas heterogeneity; K=potassium; Ni=nickel; S=sulfur; Si=silicon; V=vanadium; Zn=zinc.

^aMotorized traffic component was acquired using the principle component analysis (PCA). See Supplementary Table 6 for detailed configuration of the component. Coefficient and 95% CI were estimated by random-effects meta-analysis by cohort. Models were adjusted for parental education levels, parental countries of origin, maternal age at delivery, maternal pre-pregnancy BMI, maternal height, maternal alcohol consumption during pregnancy, maternal smoking during pregnancy, marital status, parity and age of the child at neuropsychological testing per increments of 5 ng/m³ for Cu PM_{2.5}; 100 ng/m³ for Fe PM_{2.5}; 50 ng/m³ for K PM_{2.5}; 1 ng/m³ for Ni PM_{2.5}; 200 ng/m³ for S PM_{2.5}; 100 ng/m³ for Si PM_{2.5}; 2 ng/m³ for V PM_{2.5}; 10 ng/m³ for Zn PM_{2.5} and 1 unit for motorized traffic component.

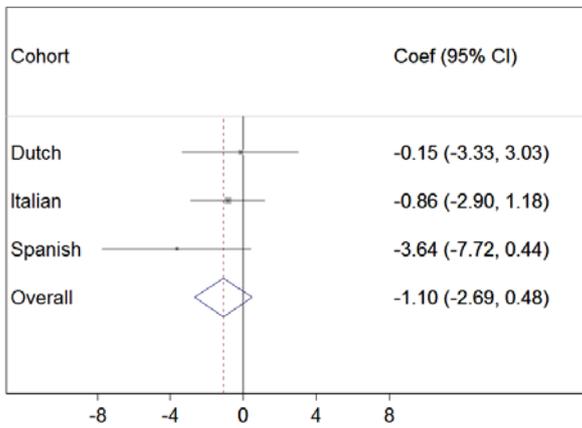
A. PM2.5 Cu (per $\Delta 5$ ng/m³)



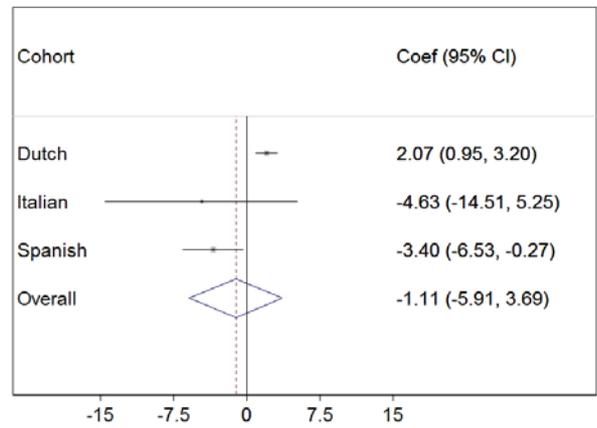
B. PM2.5 Fe (per $\Delta 100$ ng/m³)



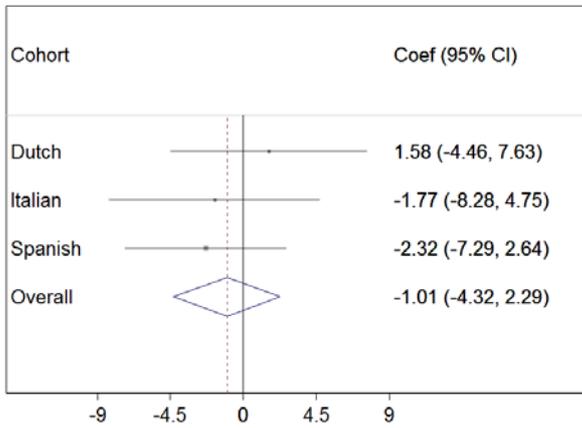
C. PM2.5 K (per $\Delta 50$ ng/m³)



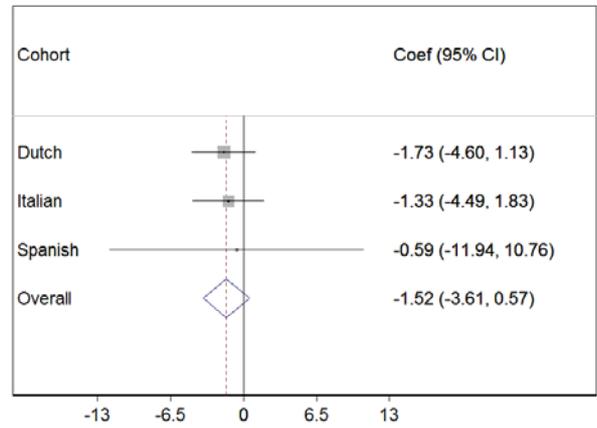
D. PM2.5 Ni (per $\Delta 1$ ng/m³)



E. PM2.5 S (per $\Delta 200$ ng/m³)



F. PM2.5 Si (per $\Delta 100$ ng/m³)



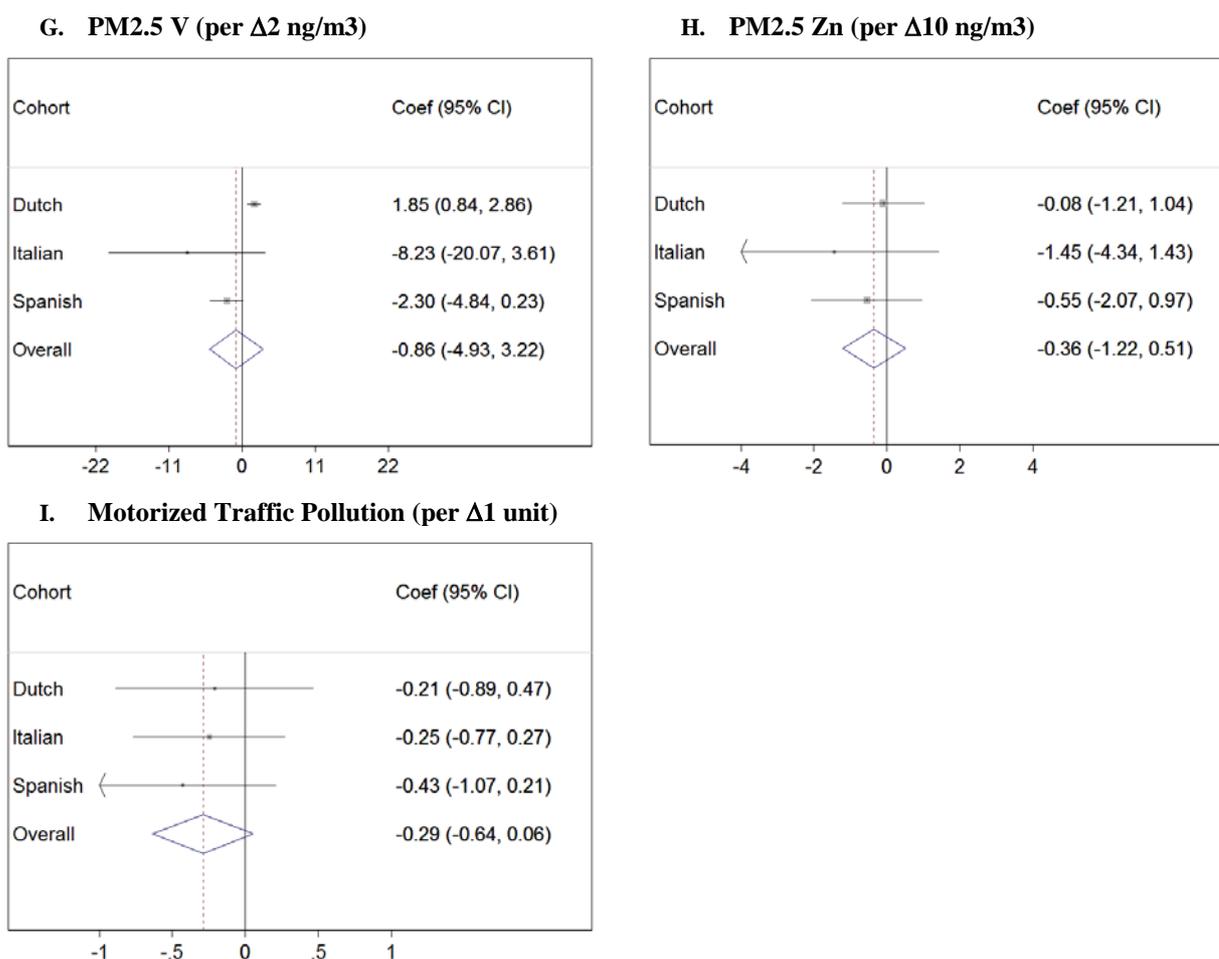


Figure 2. Fully adjusted associations of exposure to PM_{2.5} elemental composition at birth and motorized traffic pollution with fine motor function at average age of 1y in Dutch cohort, 4y in Italian cohort and 4y in Spanish cohort. Region-specific and summary risk estimates (coefficient and 95% confidence interval) for fine motor function expressed for an increase of (A) 5 ng/m³ in PM_{2.5} Cu levels, (B) 100 ng/m³ in PM_{2.5} Fe levels, (C) 50 ng/m³ in PM_{2.5} K levels, (D) 1 ng/m³ in PM_{2.5} Ni levels, (E) 200 ng/m³ in PM_{2.5} S levels, (F) 100 ng/m³ in PM_{2.5} Si levels, (G) 2 ng/m³ in PM_{2.5} V levels, (H) 10 ng/m³ in PM_{2.5} Zn levels, and (I) 1 unit in motorized traffic component levels at birth, adjusted for parental education levels, parental countries of origin, maternal age at delivery, maternal pre-pregnancy body mass index, maternal height, maternal alcohol consumption during pregnancy, maternal smoking during pregnancy, marital status, parity and age of the child at neuropsychological testing. Grey squares around region-specific coefficients represent the relative weight that the estimate contributes to the summary coefficient. Weights are from random-effects analyses. Coef= coefficient; CI=confidence intervals; Cu=copper; Fe=iron; K=potassium; Ni=nickel; S=sulfur; Si=silicon; V=vanadium; Zn=zinc.

347 **4. Discussion**

348 To our knowledge, no previous study focused on the association between exposure to
349 elemental composition of outdoor PM_{2.5} at birth and cognitive and psychomotor function in
350 childhood. This study is based on 4 European birth cohorts with data on 7,246 children.
351 Despite the lack of significant association between airborne PM_{2.5} during pregnancy and
352 cognitive and psychomotor development in childhood, we found an association with one of its
353 elemental components. Higher estimated exposure at birth to airborne iron, a main element in
354 motorized traffic pollution, was associated with lower fine motor function in children
355 assessed between 1 and 4 years of age. Exposure to elemental composition of outdoor PM_{2.5}
356 at birth was not associated with gross motor function or cognitive functions, although the
357 effect estimates of the latter were predominantly negative.

358 This study has considerable strengths: i) large sample size with western European
359 geographical extent including two countries from the northern part of Europe and two from
360 the southern part, with varying levels and sources of air pollution; ii) standardized air
361 pollution assessment which was based on validated measurements; exposure assessment of a
362 large number of elemental components measured in airborne PM_{2.5} and modelled to the
363 individual level of each participant; iii) prospective neuropsychological function assessment
364 during childhood using validated neuropsychological tests and questionnaires; iv) use of
365 advanced statistical methods including multiple imputation combined with inverse probability
366 weighting to reduce possible attrition bias in the study; and v) adjustment for various
367 socioeconomic and lifestyle variables that are known to be potentially associated with air
368 pollution exposure during pregnancy and with neuropsychological performance of the
369 offspring. However, we cannot completely discard residual confounding by
370 sociodemographic and geographic factors since adjustment for parental education levels,

371 parental country of birth, parental age, or marital status might not fully account for factors
372 that may influence cognitive and psychomotor development.

373 There are also several other limitations in our study. The neuropsychological tests and
374 the type of evaluators assessing cognitive and psychomotor functions, and the ages at which
375 children were assessed, are heterogeneous across the 4 cohorts. Nevertheless, we carefully
376 selected those tests that represent similar neuropsychological domains, adding to their
377 comparability. Another limitation of our study is related to the exposure assessment. Another
378 limitation of our study is related to the exposure assessment. Herein, the major source of
379 uncertainty lies in the inability to estimate air pollution levels at the exact period of interest
380 (i.e. at birth). Sampling campaigns were carried out between 3.5 and 9 years after the children
381 were born and historical element data from routine monitoring stations in the study areas was
382 not available to back extrapolate the levels to the period of interest. Since the temporal
383 component was missing, we assumed that the relative composition of PM_{2.5}, including the
384 relative concentration of the elements, has remained constant between births of the
385 participants and the measurements, as it has been done in a previous study on birth outcomes
386 (Pedersen et al., 2016). This assumption is based on previous studies that demonstrated spatial
387 stability over time for other traffic related air pollutants for periods stretching from 8 to 18
388 years (Gulliver et al., 2013; Cesaroni et al., 2012; Eeftens et al., 2011). Nevertheless, this
389 assumption could result in non-differential exposure misclassification which could lead to an
390 underestimation of the associations. Furthermore, we also cannot discard the possibility that
391 some of our findings occur due to chance because of the multiple comparisons performed.
392 Similar studies are necessary to confirm or refute our findings.

393 We observed a negative association between the exposure to airborne iron, the main
394 component of motorized traffic pollution, and fine motor function. Our previous published
395 study found a significant negative association between prenatal exposure to NO₂, a well-

396 known marker for traffic related air pollution, and psychomotor function assessed in children
397 between 1 and 6 years of age (Guxens et al., 2014). The association between prenatal
398 exposure to PM_{2.5} and psychomotor function was also negative, although these results were at
399 the margin of significance. Repetition of these analyses in our current study resulted in small
400 changes attributable to the changes in the study populations. Other epidemiological studies
401 also found negative associations between traffic related air pollution or some of its
402 components such as PAHs, and NO₂ and lower psychomotor function in early childhood (Xu
403 et al., 2016). This is the first study to assess a relationship between airborne iron and
404 psychomotor function. It is plausible that airborne iron is a marker for traffic related air
405 pollution and that the association that we found is in fact an association between traffic
406 related air pollution and fine motor function. However, considering that iron is a documented,
407 highly active oxidizer, and its excessive accumulation in the brain tissue can trigger
408 neuroinflammation and oxidative stress which are linked to neurodegenerative diseases,
409 neurodevelopmental disorders, and decreased cognitive function (Block et al., 2012;
410 Daugherty and Raz, 2015), we also cannot discard the possibility that the association found in
411 the current study can be attributed to the environmental exposure to airborne iron. Moreover,
412 a recent study found the presence of magnetite ultra-fine particles of external origin in human
413 brain samples. Magnetite ultra-fine particles are highly pervasive and abundant in air
414 pollution and they arise from combustion as iron-rich particulates which, upon release in the
415 air, condense and/or oxidize (Maher et al., 2016). Nevertheless, more research is needed to
416 confirm that airborne iron is one of the primary neurotoxic components of motorized traffic
417 pollution instead of a marker for a different neurotoxicant or a group of neurotoxicants
418 present in traffic related air pollution.

419 The associations between the elemental components, and the motorized traffic
420 component, and cognitive function, were predominantly negative, but significance has not

421 been reached in any case. Also in our previous published study we did not find an association
422 between prenatal exposure to NO₂ or PM_{2.5} and cognitive function (Guxens et al., 2014).
423 Postnatal exposure at the average age of 8.5 years to source apportioned elemental
424 components of outdoor PM at schools and child's working memory and attentional function at
425 corresponding time point have been assessed in a recent study which found a negative
426 relationship between exposure to source apportioned traffic pollution and the cognitive
427 functions (Basagaña et al., 2016). That study assessed specific cognitive functions such as
428 working memory and attentional function, instead of more global cognitive measurements
429 like in the current study, which might be responsible for the differing results. Also, they
430 assessed postnatal exposures in schools, as opposed to residential exposures at birth in our
431 study. Pregnancy period is of a special interest due to the relatively immature detoxification
432 mechanisms of fetuses and only partial protection of placenta and blood-brain barrier against
433 entry of environmental toxicants, and therefore higher vulnerability of the developing brain.
434 Still, brain maturation continues in childhood and adolescence and therefore a relationship
435 with postnatal exposures is plausible as well. To our knowledge, that is the only other study to
436 date that has assessed exposure to PM elements and/or source apportioned PM elements, and
437 cognitive development. Previous epidemiologic studies assessing exposure to traffic related
438 air pollutants during pregnancy and cognitive development in early childhood showed
439 conflicting results (Guxens and Sunyer, 2012; Suades-González et al., 2015).

440 In summary, we found a negative association between estimated exposure to airborne
441 iron at birth, an element highly prevalent in motorized traffic air pollution, and fine motor
442 function in childhood with a score decrease of 1.25 points for every 100 ng/m³ increase in
443 predicted iron levels at birth. Although this seemingly small decrease of 1.25% from the
444 population average might not be noticeable at an individual level, taking the population level
445 into account, this decrease will shift the distribution of fine motor performance to the left and

446 increase the number of people performing below average. Gross motor function and the
447 cognitive functions were not significantly associated with any of the PM element exposures at
448 birth, although the effect estimates of the latter were predominantly negative. Since this study
449 is the first to focus on exposure to elemental composition of outdoor PM at birth and
450 neuropsychological function in early childhood, the results require confirmation.
451 Nevertheless, they are of potential concern due to the ubiquity of traffic related air pollution,
452 which fortunately can be reduced through implementation of adequate policies worldwide.

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