

1 Radiofrequency Electromagnetic Fields from Mobile Communication: Description of Modeled Dose
2 in Brain Regions and the Body in European Children and Adolescents

3

4 Laura Ellen Birks^{1,2,3}, Luuk Van Wel⁴, Ilaria Liorni⁵, Livia Pierotti^{1,2,3}, Mònica Guxens^{1,2,3,6}, Anke
5 Huss⁴, Milena Foerster^{7,8}, Myles Capstick⁵, Marloes Eeftens^{7,8}, Hanan El Marroun^{6,9,10}, Marisa
6 Estarlich^{3,11,12}, Mara Gallastegi¹³, Llúcia González Safont^{3,11}, Wout Joseph¹⁴, Loreto Santa-
7 Marina^{3,13,15}, Arno Thielens¹⁴, Maties Torrent³, Tanja Vrijkotte¹⁶, Joe Wiart¹⁷, Martin Röösl^{7,8},
8 Elisabeth Cardis^{1,2,3}, Roel Vermeulen^{4,18,19}, Martine Vrijheid^{1,2,3}

9 ¹ISGlobal, Barcelona, Spain

10 ²Pompeu Fabra University, Barcelona, Spain

11 ³Spanish Consortium for Research on Epidemiology and Public Health (CIBERESP), Instituto de
12 Salud Carlos III, Madrid, Spain

13 ⁴Institute for Risk Assessment Sciences (IRAS), Utrecht University, Utrecht, The Netherlands

14 ⁵Foundation for Research on Information Technologies in Society (IT'IS), Zurich, Switzerland

15 ⁶Department of Child and Adolescent Psychiatry/Psychology, Erasmus MC, University Medical
16 Centre, Rotterdam, The Netherlands

17 ⁷Swiss Tropical and Public Health Institute, Basel, Switzerland

18 ⁸University of Basel, Basel, Switzerland

19 ⁹ Department of Pediatrics, Erasmus MC

20 ¹⁰ Department of Psychology, Education and Child Studies – Erasmus University Rotterdam

21 ¹¹Epidemiology and Environmental Health Joint Research Unit, FISABIO-Universitat Jaume I-
22 Universitat de València, Valencia, Spain

23 ¹² Faculty of Nursing and Chiropody, Universitat de València, Spain

24 ¹³BIODONOSTIA Health Research Institute, Dr. Begiristain Pasealekua, San Sebastian, Spain

25 ¹⁴Department of Information Technology, Ghent University/IMEC, Technologiepark 126, Ghent,
26 9052, Belgium

27 ¹⁵Department of Health of the Basque Government, Public Health division of Gipuzkoa, Donostia-
28 San Sebastián, Spain

29 ¹⁶Department of Public Health, Amsterdam UMC, University of Amsterdam, Amsterdam Public
30 Health Research Institute, Amsterdam, the Netherlands

31 ¹⁷Télécom ParisTech, LTCI University Paris Saclay, Chair C2M, Paris, France

32 ¹⁸Julius Center for health Sciences and Primary Care, University Medical Center Utrecht

33 ¹⁹School of Public Health, Imperial College London, London, UK

34 **CORRESPONDING AUTHOR**

35 Martine Vrijheid, ISGlobal-Campus Mar, Doctor Aiguader, 88, 08003 Barcelona

36 Tel. +34 932 147 346 | Fax +34 932 147 301. Email: martine.vrijheid@isglobal.org

37 **Short running title: Child and Adolescent RF dose in Europe**

38 **Conflict of interest: none declared**

39 **Keywords:** Cell Phone Use, Children’s Health, Electromagnetic Fields, Radio Waves, Temporal
40 Lobe, Frontal Lobe

41 **Sources of financial support:**

42 **GERoNiMO project:** This work is supported by the European Union (grant 603794).

43 **HERMES project:** This work is supported by the Swiss National Science Foundation (project
44 number 138190). This research is also supported by the Swiss Research Foundation for Electricity
45 and Mobile Communication (reference number 41).

46 **ABCD, The Netherlands:** This work is supported by the Netherlands Organization for Health
47 Research and Development (grant 2100.0076) and within the programme Electromagnetic Fields
48 and Health Research (grants 85600004 and 85800001).

49 **Generation R:** The general design of the Generation R Study is made possible by financial support
50 from the Erasmus Medical Center, Rotterdam, the Erasmus University Rotterdam, the Netherlands
51 Organization for Health Research and Development (ZonMw), the Netherlands Organization for
52 Scientific Research (NWO), and the Ministry of Health, Welfare and Sport. Hanan El Marroun is
53 supported by the European Union’s Horizon 2020 research and innovation program (no.733206
54 LifeCycle), Stichting Volksbond Rotterdam and the Dutch Brain Foundation (De Hersenstichting,
55 project number GH2016.2.01).

56 **INMA, Gipuzkoa:** This study was funded by grants from Instituto de Salud Carlos III (FIS-
57 PI13/02187), CIBERESP, Department of Health of the Basque Government (2015111065), and the
58 Provincial Government of Gipuzkoa (DFG15/221) and annual agreements with the municipalities
59 of the study area.

60 **INMA, Menorca:** This study was funded by grants from Instituto de Salud Carlos III (Red INMA
61 G03/176; CB06/02/0041; 97/0588; 00/0021-2; PI061756; PS0901958; PI14/00677 incl. FEDER
62 funds), CIBERESP, Beca de la IV convocatoria de Ayudas a la Investigación en Enfermedades
63 Neurodegenerativas de La Caixa, and EC Contract No. QLK4-CT-2000-00263.

64 **INMA, Sabadell:** This study was funded by grants from Instituto de Salud Carlos III (Red INMA
65 G03/176; CB06/02/0041; PI041436; PI081151 incl. FEDER funds; PI12/01890 incl. FEDER funds;
66 CP13/00054 incl. FEDER funds, MS13/00054, CP18/00018), CIBERESP, Generalitat de Catalunya-
67 CIRIT 1999SGR 00241, Generalitat de Catalunya-AGAUR (2009 SGR 501, 2014 SGR 822), Fundació
68 La maratón de TV3 (090430), Spanish Ministry of Economy and Competitiveness (SAF2012-
69 32991 incl. FEDER funds), Agence Nationale de Sécurité Sanitaire de l'Alimentation de
70 l'Environnement et du Travail (1262C0010, EST-2016 RF-21), EU Commission (261357, 308333 and
71 603794). We also acknowledge support from the Spanish Ministry of Science and Innovation and
72 and State Research Agency through the "Centro de Excelencia Severo Ochoa 2019-2023" Program
73 (CEX2018-000806-S), and support from the Generalitat de Catalunya through the CERCA Program.

74 **INMA, Valencia:** This study was funded by Grants from UE (FP7-ENV-2011 cod 282957 and
75 HEALTH.2010.2.4.5-1), Spain: ISCIII (G03/176; FIS-FEDER: PI11/01007, PI11/02591, PI11/02038,
76 PI13/1944, PI13/2032, PI14/00891, PI14/01687, PI16/1288, and PI17/00663; Miguel Servet-FEDER
77 CP11/00178, CP15/00025, and CPII16/00051), and Generalitat Valenciana: FISABIO (UGP 15-230,
78 UGP-15-244, and UGP-15-249).

79 **Acknowledgements:** The authors would particularly like to thank all participants for their
80 generous collaboration.

81

82 **Abstract**

83 Background: Little is known about radiofrequency electromagnetic fields (RF) from mobile
84 technology and resulting dose in young people. We describe modeled integrated RF dose in
85 European children and adolescents combining own mobile device use and surrounding sources.

86 Methods: Using an integrated RF model, we estimated the daily RF dose in the brain (whole-brain,
87 cerebellum, frontal lobe, midbrain, occipital lobe, parietal lobe, temporal lobes) and the whole-
88 body in 8,358 children (ages 8-12) and adolescents (ages 14-18) from the Netherlands, Spain, and
89 Switzerland during 2012-2016. The integrated model estimated RF dose from near-field sources
90 (digital enhanced communication technology (DECT) phone, mobile phone, tablet, and laptop) and
91 far-field, surrounding, sources (mobile phone base stations via 3D-radiowave modeling or RF
92 measurements).

93 Results: Adolescents were more frequent mobile phone users and experienced higher modeled RF
94 doses in the whole-brain (median 330.4 mJ/kg/day) compared to children (median 81.8
95 mJ/kg/day). Children spent more time using tablets or laptops compared to adolescents, resulting
96 in higher RF doses in the whole-body (median whole-body dose of 81.8 mJ/kg/day) compared to
97 adolescents (41.9 mJ/kg/day). Among brain regions, temporal lobes received the highest RF dose
98 (medians of 274.9 and 1,786.5 mJ/kg/day in children and adolescents, respectively) followed by
99 the frontal lobe. In most children and adolescents, calling on 2G networks was the main
100 contributor to RF dose in the whole-brain (medians of 31.1 and 273.7 mJ/kg/day, respectively).

101 Conclusion: This first large study of RF dose to the brain and body of children and adolescents,
102 shows that mobile phone calls on 2G networks are the main determinants of brain dose, especially
103 in temporal and frontal lobes, whereas whole-body doses were mostly determined by tablet and
104 laptop use. The modeling of RF doses provides valuable input to epidemiological research and to
105 potential risk management regarding RF exposure in young people.

106

107 **1. Introduction**

108 Over the past thirty years, mobile communication technology has transformed society, providing
109 new platforms for near-constant communication, media and entertainment consumption, and
110 socializing. This means that children today will experience more radiofrequency electromagnetic
111 fields (RF) exposure in childhood and a higher accumulated lifetime exposure, compared to
112 previous generations (Markov and Grigoriev 2015; Rosenberg 2013; Otto and von Mühlendahl
113 2007). There is concern that elevated exposure to RF at a young age, while organs are rapidly
114 developing, could lead to adverse health effects in childhood or later in life (Rice and Barone 2000;
115 Feychting 2011; Kheifets et al. 2005; Schüz 2005). Epidemiological research has yet to
116 comprehensively describe these recent levels of RF exposure in young populations, which is
117 necessary for understanding its possible long- and short-term health effects. Therefore, a clearer
118 understanding of RF exposure, specifically in children and adolescents, is urgently needed during
119 this new era in mobile communication.

120

121 Epidemiological studies estimating children's RF exposure have been rare, and the recent studies
122 that have attempted to describe RF exposure in children have their limitations. For example,
123 epidemiological studies that estimated exposure based solely on mobile phone use (Sudan et al.
124 2016; Abramson et al. 2009) or solely on geospatial modeling (Beekhuizen et al. 2013; Schoeni et
125 al. 2016) do not account for RF doses from far- and near-field sources together. In studies that
126 accounted for far- and near-field sources together, measurements could not account for the
127 differences in RF exposure from different types of device use activities (such as calling, texting,
128 internet browsing, or video streaming) or network coverage (2G, 3G, or WiFi), factors which effect

129 device output power, and thus RF dose received (Huss et al. 2015; Guxens et al. 2016).
130 Alternatively, studies measuring personal RF exposure are sometimes in small populations
131 ($n < 1,000$), as they are labor intensive, and are unable to account for dose to specific tissues of the
132 body, as measured values depend highly on the distance between the emitting source and the
133 measurement device, which is not necessarily the same as the distance between the emitting
134 source and the body (Birks et al. 2018; Roser et al. 2017; Calvente et al. 2015; Thomas et al. 2010;
135 2009; 2008).

136 A recent study in Switzerland used dose models which integrated near- and far-field estimates,
137 types of device use activities, and network coverage to estimate RF dose to the whole-brain, the
138 brain's gray matter, and the whole-body (Lauer et al. 2013; Roser et al. 2015). However, these
139 analyses did not model dose in specific regions of the brain and were limited to small samples
140 (Schoeni, Roser, and Rösli 2015; Foerster et al. 2018; Schoeni, Roser, and Rösli 2017). Therefore,
141 further estimates of modeled RF dose in specific regions of the brain in larger populations are
142 necessary.

143 This study aims to model RF dose in 7 regions of the brain and whole-body, describe this modeled
144 dose and the mobile device use habits contributing to it, in two age groups from a large sample of
145 children (ages 8-12) and adolescents (ages 14-18) across three European countries between 2012
146 and 2016. This work combines existing data from four population based cohort studies, resulting
147 in the largest epidemiological RF dose description of children's brains and bodies to date.

148 **2. Methods**

149 *2.1 Study design and population*

150 As part of the Generalized EMF Research using Novel Methods (GERoNiMO) Project
151 (<http://radiation.isglobal.org/geronimo>) and the Radiofrequency electromagnetic fields exposure
152 and brain development (REMBRANDT) Project ([http://radiation.isglobal.org/index.php/
153 nl/radiation-programme-projects/2018-06-19-08-32-04/rembrandt](http://radiation.isglobal.org/index.php/nl/radiation-programme-projects/2018-06-19-08-32-04/rembrandt)), four population-based
154 prospective cohorts spanning Europe (Table 1) were combined for analysis regarding mobile
155 device use at ages 8-18. These were: the Dutch Amsterdam Born Children and their Development
156 Study (ABCD) (Eijsden et al. 2011), the Dutch Generation R Study (Generation R) (Kooijman et al.
157 2016), Switzerland's Health Effects Related to Mobile phone use in adolescentS (HERMES)
158 (Schoeni, Roser, and Rössli 2015), and the Spanish Environment and Childhood Project (INMA)
159 (Guxens et al. 2012). Adolescents in HERMES were recruited for a one year follow-up study with
160 two data collection periods: at baseline and one year later. In this analysis, we used HERMES data
161 collected at baseline. The Spanish INMA cohort consisted of several regional subcohorts,
162 comprising of Gipuzkoa, Sabadell, Valencia, (collectively referred to as INMA young) and Menorca
163 (referred to as INMA Menorca). Enrollment in the ABCD, Generation R, and INMA occurred during
164 the mother's pregnancy, spanning years 1996-2008; while adolescents in the HERMES cohort were
165 recruited at ages 13-14 between 2012 and 2014. In all cohorts, informed consent was obtained
166 from all participants' parents or guardians in accordance with each center's institutional review
167 board or ethics committee. Data regarding mobile device use was collected at different ages
168 during 2012 to 2016, depending on the cohort (Table 1). Across all cohorts, 8,358 participants met
169 our inclusion criteria of having information on mobile device use and proximity to mobile phone
170 base stations at ages 8-18. Excluded participants that were missing this information were either
171 lost to follow-up or had incomplete questionnaires. Based on time of data collection for this
172 analysis, cohorts were grouped into two age groups: children (ABCD, Generation R, and INMA
173 young) and adolescents (HERMES and INMA Menorca).

174 *2.2 Exposure to RF sources and modeling of RF dose to brain and body*

175 *Use of mobile communication devices*

176 In ABCD, Generation R, and INMA young cohorts, parents were asked via questionnaire to
177 estimate children's frequency and duration of mobile device use when children were 8-12 years
178 old. Adolescents from the HERMES and INMA Menorca reported this information themselves via
179 questionnaire (Table 1) when they were 14 and 18 years old, respectively. The following variables
180 were collected regarding use (yes/no) and average daily or weekly duration of use: Digital
181 Enhanced Cordless Telecommunications (DECT) calls, mobile phone calls, mobile phone internet
182 browsing, mobile phone emailing, laptop use, tablet use, and laptop or tablet connection to
183 internet. Questionnaires asked for number of text messages or app-based messages sent.
184 Questionnaires also asked for model of mobile phone (smartphone, bar, slide, or flip), except in
185 ABCD. In HERMES, adolescents were asked to report "daily data traffic" on mobile phone instead
186 of browsing or emailing, which we have reported as mobile phone internet browsing. HERMES
187 adolescents were also asked to report tablet and laptop use together, which we have reported as
188 tablet use. In HERMES and INMA Menorca, adolescents were asked to report laterality (left or
189 right) of mobile phone use, and use of Bluetooth or other hands-free device during mobile phone
190 calls. For questionnaires, see supplemental materials (Supplementary Table S1).

191 *Far-field RF from mobile phone base stations in the home*

192 For most participants, daily RF exposure in the home from nearby mobile phone base stations was
193 estimated using a three-dimensional radiowave propagation model, NISMap. NISMap models
194 estimated RF exposure from base stations emitting the following downlink frequencies: 800 MHz
195 (in INMA young and INMA Menorca only), 900 MHz, 1800 MHz, and 2100 MHz. Based on
196 geocoding of the participant's home address and the floor level of the participant's bedroom,

197 NISMap estimated RF exposure in power density (mW/m^2), considering the three dimensional
198 environment by including topography and detailed information about nearby mobile phone base
199 stations (such as output power, height above ground and direction) (Bürge et al. 2010). Information
200 regarding characteristics of the participant's bedroom was collected via questionnaires along with
201 device use details. For questionnaires, see Supplementary Table S1. In HERMES, additional daily
202 exposure at school from mobile phone base stations was modeled with NISMap based on the
203 adolescent's geocoded school address.

204 *Far-field RF from other sources in all microenvironments*

205 To estimate far-field exposure from other sources (such as FM, TV, uplink, DECT, and WiFi sources)
206 in all microenvironments, estimates were made using personal RF exposimeter measurements
207 taken over up to 72 hours in a previous study in a subset of cohort participants in all regions (Birks
208 et al. 2018; Eeftens et al. 2018) (Supplementary Table S2). From this point on, we will refer to
209 these measurements as exposimeter estimates. Exposimeter estimates from the previous study
210 and time spent in each microenvironment (home, school, outdoors, traveling) were averaged for
211 each microenvironment and frequency band and matched by regional cohort. However, the
212 Generation R cohort was not part of this previous study, therefore exposimeter measurements
213 taken in ABCD were used for estimates in both ABCD and Generation R, given the similarities in
214 infrastructure and population density of the studies' centers: Amsterdam and Rotterdam. For a
215 description of exposimeter estimates, please see Supplementary Table S3.

216 For 114 children in INMA young and 21 adolescents in INMA Menorca, it was not possible to
217 geocode home addresses to use for NISMap calculations. For this reason, regional medians of
218 exposimeter estimates were used to estimate far-field RF exposure from mobile base stations in

219 the home (Supplementary Table S2). For far-field RF estimates used in the home, see
220 Supplementary Table S3.

221 *Integrated model of RF dose in the brain and body*

222 To model RF dose in specific regions and tissues of the brain and body, Liorni et al. (Liorni et al.
223 2020) developed an integrated dose model to include many relevant RF sources, based on specific
224 absorption rate (SAR) transfer approximations developed and applied in van Wel et al. in 2018
225 (vanWel et al. 2020) on the basis of the data collected within the Characterization of exposure to
226 RF induced by new uses and technologies of mobile communication systems (CREST) project
227 (<https://www.isglobal.org/en/-/crest>) and the previously mentioned GERoNiMO project. This
228 integrated model estimates the exposure to RF systems in the near-field and far-field, allowing for
229 a broad exposure assessment. The model takes into account source specific attributes (source
230 type, output power, operating frequency), personal characteristics (body mass, weight), and the
231 specific exposure scenario (position relative to the body, type of use, duration of use), allowing for
232 better dose estimation and insight in the contribution of different sources and uses to the total RF
233 dose received (Cabr -Riera, Marroun, et al. 2020).

234 SAR approximations were derived on the basis of a large-scale numerical study analysing the
235 exposure of the human anatomical phantom belonging to the Virtual Population (Gosselin et al.
236 2014) developed by the IT'IS Foundation (<https://itis.swiss>) to different RF systems in several
237 exposure scenarios. Virtual population phantoms were assigned to participants based on age and
238 mass of the participant (Supplementary Table S4). SAR values were transformed to dose values by
239 multiplying the SAR by output power of the device then by relevant exposure durations. The
240 average output power of these devices was derived from the literature and expert opinion

241 (Persson et al. 2012; Joseph et al. 2013). Output powers for device use specific to device, network,
242 and activity can be found in Supplementary Table S5.

243 RF dose in 7 brain regions (whole-brain, cerebellum, frontal lobe, midbrain, parietal lobe, occipital
244 lobe, and temporal lobes) and the whole-body was calculated by summing contributions of all
245 near- and far-field exposure scenarios, resulting in region-specific doses in mJ/kg/day
246 (Supplementary Table S6). The model calculated the integrative dose from all sources combined
247 and the relative contribution of each source. For a summary of assumptions or estimates used in
248 exposure modeling, we refer to Supplementary Methods S1 and Table S7.

249 *2.3 Sociodemographic factors*

250 Sociodemographic factors were collected in all cohorts via questionnaires during follow-up or at
251 baseline (Table 2). As few previous studies have investigated sociodemographic factors and
252 potential associations with mobile device use and RF exposure (Langer et al. 2017; Birks et al.
253 2018; Andone et al. 2016), we considered sociodemographic factors in our analyses to further
254 explore possible associations. Characteristics of children such as age and sex and characteristics of
255 mothers such as age at birth, marital status (living with a partner or living alone), education
256 (highest level completed: primary, secondary, university or higher), and parity (0 children, 1 child,
257 >1 child) were considered.

258 *2.5 Statistical analysis*

259 Among participants with at least one variable available on mobile device use (n=8,358), we
260 performed multiple imputation of missing participant characteristics and duration of device use
261 values using chained equations where 25 completed datasets were generated and analyzed using
262 the standard combination rules for multiple imputation (Graham, Olchowski, and Gilreath 2007;

263 Sterne et al. 2009). Distributions in the imputed datasets were very similar to those in the original
264 data set (data not shown).

265 The integrated dose model could not be applied to imputed datasets. Therefore, missing values
266 necessary for use of the RF dose model (sex, age, height, weight, duration of device activities,
267 proportion of right/left use in HERMES and INMA Menorca, proportion of 2G/3G use in HERMES)
268 were replaced using the mean value for the individual of those variables from the 25 imputed
269 datasets per cohort (Supplementary Table S7). These datasets were used for calculation of the
270 dose in the model.

271 Daily median device use durations, individual characteristics, and sociodemographic factors were
272 described by cohort. Total daily median RF dose (mJ/kg) was modeled and described in each
273 cohort for the whole-brain, cerebellum, frontal lobe, midbrain, occipital lobe, parietal lobe,
274 temporal lobes, and the whole-body. Median modeled dose was described for relevant brain
275 regions with respect to the following specific RF sources: all sources combined, DECT calls, 2G
276 calls, 3G calls, mobile phone data use, laptops, tablets, and far-field.

277 Associations between sociodemographic factors with log-transformed RF dose to the whole-brain
278 and whole-body were estimated using mixed models with random cohort effects. Cohort random
279 effects allowed models to capture age effects and other exposure-relevant factors that are
280 clustered per cohort. Geometric mean ratios and 95% confidence intervals were calculated.
281 Models for the association between sociodemographic factors and log-transformed RF dose were
282 adjusted for sex but not for age, given the high correlation between age and cohort.

283 All analyses were performed using Stata 14 statistical software (Stata Corporation, College Station,
284 Texas) and R statistical software (R Core Team 2013). Data from all cohorts was sent to and
285 analyzed at ISGlobal in Barcelona, Spain.

286 **3. Results**

287 *Children and adolescents' characteristics and device use habits*

288 In children of ABCD (mean age 12 years), Generation R (10 years), and INMA young (9 years), the
289 prevalence of any mobile phone use was 82%, 50%, and 10%, respectively. In HERMES (14 years)
290 and INMA Menorca (18 years), adolescents were more prevalent mobile phone users with 95%
291 and 99% reporting any use of a mobile phone, respectively (Table 2). While adolescents were
292 more prevalent users of mobile phones than children, the type of use differed among cohorts.
293 Adolescents in HERMES spent more time calling, with a daily median of 6 minutes for mobile
294 phone and 3 minutes for DECT calls, while adolescents in INMA Menorca were more frequent
295 users of text messaging or mobile app messaging (median of 73 messages per day). In child
296 cohorts, however, these children spent more time using tablets than adolescents with ABCD,
297 Generation R, and INMA young cohorts reporting median values of 30, 30, and 12 minutes per day,
298 respectively. For a detailed description of device use in all cohorts, see Table 2.

299 *Modeled daily RF dose in regions of the brain*

300 Overall, modeled median RF dose in the whole-brain was 91.7 mJ/kg/day (Table 3), though this
301 varied widely between cohorts and age groups (83.7 mJ/kg/day in children and 330.4 mJ/kg/day in
302 adolescents) (Supplementary Figure S1a). Models estimated that temporal lobes received more RF
303 than other brain regions (medians 274.9 and 1,786.5 mJ/kg/day in children and adolescents,
304 respectively), followed by the frontal lobe (123.4 and 582.9 mJ/kg/day in children and adolescents,
305 respectively) (Figures 1a-b). In children, most other regions, including the cerebellum, the
306 midbrain, the occipital lobe, and the parietal lobe received a dose of less than 102 mJ/kg/day,
307 while in adolescents these brain regions received a dose of less than 300 mJ/kg/day.

308 *Contributions by RF sources to modeled daily RF dose in regions of the brain*

309 Contributors to modeled daily RF dose in the brain varied by age groups due to different habits of
310 device use. In children, modeled dose in the whole-brain was comprised of RF from 2G (31.1
311 mJ/kg/day), far-field (8.5 mJ/kg/day) and tablet use (6.7 mJ/kg/day), with dose in the temporal
312 lobes dominated by 2G calls. These results were mainly driven by ABCD and Generation R, since
313 among the INMA young children, whole-brain dose was mostly comprised of exposure from far-
314 field (10.5 mJ/kg/day) exposure. Modeled dose in the whole-brain, frontal and temporal lobes in
315 adolescents overwhelmingly originated from mobile phone calls on 2G networks (median 2G dose
316 was 273.7 mJ/kg/day for the whole-brain and 1,595.9 mJ/kg/day for the temporal lobe)
317 (Supplementary Table S8) (Figures 2a-c). The remaining dose in the brains of adolescents was very
318 low and mostly comprised of RF from 3G (7.3 mJ/kg/day), DECT (6.1 mJ/kg/day), farifield (5.9
319 mJ/kg/day), and mobile phone data (0.5 mJ/kg/day).

320 *Modeled daily RF dose and its contributors in the whole-body*

321 Daily median whole-body dose in children (81.8 mJ/kg) resulted mainly from tablet use (31.2
322 mJ/kg/day) (Table 3). Meanwhile, median daily whole-body dose in adolescents (41.9 mJ/kg)
323 resulted mainly from 2G and far-field exposure (5.2 and 4.7 mJ/kg/day, respectively), though this
324 varied between adolescent cohorts (Figure 2d).

325 *Sociodemographic characteristics and RF dose*

326 Older children (per year of age) and females experienced higher modeled RF dose in the whole-
327 brain (Supplementary Table S9). Children of less educated mothers and of mothers that lived alone
328 received more RF dose in the whole-brain. Children of younger mothers had slightly higher RF

329 dose in the brain. Whole-body RF dose was higher in males and children or adolescents of mothers
330 that were more educated (Supplementary Table S9).

331 **4. Discussion**

332 In this study, we modeled daily RF dose in 7 brain regions and the whole-body in children and
333 adolescents from four European prospective cohorts, based on parent- or self-reported mobile
334 device use and modeled or measured exposure to far-field RF sources. Adolescents were more
335 prevalent mobile phone users, and therefore experienced much higher RF dose to the brain than
336 children. Children spent relatively more time using tablets or laptops than calling, and experienced
337 higher RF dose to the whole-body compared to adolescents. Both children and adolescents
338 received the highest RF dose in the temporal and frontal lobes of the brain. Mobile phone calling
339 on 2G networks was the main contributor to frontal and temporal lobe dose in both age groups,
340 followed by far-field exposure in children, and followed by calling on 3G or DECT networks in
341 adolescents.

342 This is the first study to quantify modeled RF dose in the whole-body and specific regions of the
343 brain (whole-brain, cerebellum, frontal lobe, midbrain, parietal lobe, occipital lobe, and temporal
344 lobes) in children and adolescents and to describe how this dose differs with RF sources. Previous
345 modeling of RF dose in the whole-brain in adolescents has been done in Switzerland in the
346 HERMES cohort. In 2015, Roser et al. estimated mean RF doses in the whole-brain (1,559.7
347 mJ/kg/day) in a sample of 400+ adolescents (Roser et al. 2015). Alternatively, Foerster et al. 2018
348 used average reported device use values over one year (at baseline and follow-up) to apply an
349 exposure estimate model to a HERMES sample (n=676) and estimated mean RF dose in the whole-
350 brain (858 mJ/kg/day) (Foerster et al. 2018). For both of those analyses, RF dose estimates were
351 higher than means in our analysis of HERMES (data not shown). This is likely due to differences in

352 assumptions in the exposure model. For example, Foerster et al. accounted for mobile phone data
353 traffic use on mobile phone networks, whereas our calculations assumed mobile phone data use
354 on WiFi networks. Foerster et al. also accounted for daily duration of carrying the mobile phone
355 near the body, an exposure scenario not accounted for in our analysis as this variable was not
356 available in all cohorts.

357 Children's and adolescents' temporal lobes and frontal lobes received more RF dose than all other
358 regions of the brain. This makes sense considering where the mobile phone is held (next to the ear
359 or in front of the face) and where tablets are held (in front of the face). The frontal lobe is
360 important for various higher-order cognitive functions, such as managing emotions, attentional
361 control, abstract thinking, among others (Rosso et al. 2004; Baars and Gage 2010). Meanwhile,
362 functions of the temporal lobes are involved with creating and storing new memories, language
363 recognition, among others (Baars and Gage 2010). Previous epidemiological studies evaluating the
364 association between RF dose in the brain and neurodevelopmental outcomes in children are few
365 and specific to HERMES adolescents in Switzerland (Schoeni, Roser, and Rössli 2015; Foerster et
366 al. 2018). In these studies, researchers found RF dose to the whole-brain was associated with
367 decrease in figural memory performance. Further analysis of RF dose in the frontal and temporal
368 lobes and cognitive function and behavioral outcomes is needed to evaluate the potential long-
369 and short-term consequences of RF dose levels and neurocognitive development.

370 In children and adolescents who reported any mobile phone calling, 2G networks (including
371 General Packet Radio Service (GPRS) or Enhanced Data rates for Global Evolution (EDGE)
372 networks) contributed most to modeled daily RF dose in the whole-brain, frontal lobes, and
373 temporal lobes. In our analysis, the proportion of 2G network usage was individually imputed for
374 the HERMES cohort based on objective service provider data for n=322, while for other cohorts,

375 we used a country-wide average for all children based on a study completed (with objective data
376 from a software application on individual mobile phones) around the same years that children
377 reported device use (Langer et al. 2017; Goedhart et al. 2018). We must acknowledge that RF dose
378 estimates for children and adolescents that made mobile phone calls could vary widely depending
379 on 2G proportions used in the model. However, given the output power of a mobile phone while
380 calling on a 2G network (89.7 mW) is 200 times that of that on a 3G network (0.45 mW), any
381 proportion of calling on a 2G network would dominate RF exposure to the brain. Therefore, our
382 modeled dose results demonstrate that RF dose to the brain could be greatly reduced by
383 avoidance of mobile phone calls on 2G networks.

384 Generally, our results illustrate changing exposure as children mature and begin to use their own
385 devices. We see that in child cohorts, especially the youngest (INMA Young), RF dose to the brain
386 was very low and comprised of mostly far-field sources; while in older children and adolescents, RF
387 dose was higher and dominated by 2G exposure from calling on their own devices. On a more
388 granular level, we found that RF dose in the brain varied with child or adolescent characteristics
389 and sociodemographic factors. Particularly, females, older children, children of mothers living
390 alone, and children of mothers with less education were more prevalent mobile phone users and
391 therefore experienced higher RF dose in the brain. Previous research has also demonstrated more
392 prevalent mobile phone use in females and older children (Langer et al. 2017; Andone et al. 2016).
393 Langer et al. demonstrated that adolescents whose mothers had less education made more
394 frequent and longer mobile phone calls and used more mobile phone data (Langer et al. 2017),
395 while Birks et al. illustrated that personal environmental RF was higher in children with less
396 educated mothers (Birks et al. 2018). However, Langer's and Birks' analyses did not investigate
397 mobile phone use or environmental RF and maternal marital status. Therefore, our findings add to

398 current evidence pointing towards a complex relationship between mobile device use and
399 sociodemographic factors.

400 While adolescents were more frequent users of mobile phones in our analysis, children were
401 instead more frequent users of laptops or tablets. Estimates show this led to a higher RF dose in
402 the whole-body in children than in adolescents. RF dose in the whole-body was higher in boys and
403 children of mothers with more education. RF dose to the whole-body in small children has not yet
404 been studied, to our knowledge, and its possible associations with health outcomes should be
405 investigated in the future.

406 Our study has some important strengths, including its relatively large sample size and wide age
407 range across three countries, and the harmonized and detailed information regarding mobile
408 device use, individual characteristics, as well as sociodemographic factors. To date, this is the first
409 study to model RF dose in specific regions (whole-brain, cerebellum, frontal lobe, midbrain,
410 parietal lobe, occipital lobe, and temporal lobes) of the brain as a result of mobile device use and
411 exposure to far-field RF sources together, in children and adolescents. To model this dose, we
412 have used an integrated dose model, one of the most comprehensive RF dose estimation tools
413 available in epidemiological research. Questionnaire data used for this modeling accounted for
414 very detailed information regarding mobile device use. While previous studies may have estimated
415 RF exposure based solely on mobile phone calls or proximity to nearby mobile phone base
416 stations, our study was able to combine these exposure sources to model an integrated RF dose.

417 Our study also has several limitations. RF dose estimates are based on detailed modeling for which
418 not all input data was available in our study population. Therefore in some cases, assumptions had
419 to be made regarding several factors: laterality of mobile phone use, 2G/3G network use
420 proportions, activities while using mobile phone data or laptops and tablets, WiFi data transfer

421 rates (54 Mbps), time spent in certain microenvironments, and RF exposure from other far-field
422 sources (Supplementary Methods S1 and Table S7). However, our exposure modeling was recently
423 used by Calbré-Riera et al and their sensitivity analyses explored additional higher-exposure and
424 lower-exposure assumptions (Cabré-Riera, vanWel, et al. 2020). Results demonstrated that
425 changes in assumptions led to only marginal RF dose variations. Nevertheless, future
426 epidemiological studies wishing to model RF dose should aim to collect as much information as
427 possible on these factors in order to avoid these assumptions.

428 Additionally, previous research has found that adolescents and young people typically
429 overestimate their duration of mobile phone calling, compared to objectively recorded measures
430 (via software modified phones or service provider records) (Foerster et al. 2018; Langer et al.
431 2017). However, the self-reported estimates have been shown to accurately distinguish high-
432 frequency users from low-frequency users (Langer et al. 2017). Regarding parents' estimation of
433 child's device use, there are no studies validating parental reporting. Furthermore, there are no
434 studies validating self or parental reporting of tablet or laptop use. It remains a possibility that
435 parents under or overestimated tablet and laptop use, from which models estimated the most RF
436 dose in the whole-body. However, in large scale population-based settings it would not be feasible
437 to objectively monitor mobile phone, tablet, or laptop use. While estimating far-field exposures to
438 RF, our modeling combined individual NISMap estimates with regional exposimeter
439 measurements. While this method combines measurements with systematic differences, we felt it
440 was important to try to capture many far-field exposures (as captured by exposimeter
441 measurements) together with mobile phone base stations (as modeled by NISMap). Finally, the
442 population included in this study was from population-based cohort studies. Some participants in
443 these cohort studies were lost to follow-up or had incomplete questionnaires (no data was
444 available regarding mobile device use) at ages 8-18 , meaning they had to be excluded from

445 analysis in this study. With this possibility for selection bias, our results may not represent the
446 general population, limiting the external validity of our results (Szklo 1998).

447 **5. Conclusion**

448 Our study estimates for the first time in a large sample of children and adolescents RF dose in the
449 whole-brain and whole-body, which regions of the brain receive the highest RF dose from mobile
450 communication, and which sources and devices are the most relevant contributors in these age
451 groups. Brain doses, especially those in the temporal and frontal lobes, were predominantly
452 determined by mobile phone calls on 2G networks and less by other RF sources. The modeling of
453 RF doses through use of an integrated dose model is a useful tool for future epidemiological
454 research and potential risk management regarding RF exposure.

455

456

Table 1: Description of cohorts in analysis

	Cohort	Location	Enrollment		Data collection of mobile device use			
			Time period	N	Mean age	Reporter	Time period	N
Children	ABCD	Amsterdam, NL	2003-2004	8,266	12 y	parent	2015-2016	2,593
	Generation R	Rotterdam, NL	2002-2006	9,901	10 y	parent	2015-2016	3,304
	INMA young	Gipuzkoa, Sabadell, and Valencia, ES	2003-2008	2,271	8-10 y	parent	2014-2016	1,311
Adolescents	HERMES	Switzerland, CH	2012-2014	892	14 y	self	2012-2014	892
	INMA Menorca	Menorca, ES	1997-1998	482	18 y	self	2015-2016	258
								25,732
								8,358

Abbreviations: ABCD, Amsterdam Born Children and their Development ; CH, Switzerland; ES, Spain; HERMES, Health Effects Related to Mobile phone use in adolescents; INMA, Spanish Childhood and Environment Project; NL, the Netherlands; y, years.

457

458

459

Table 2: Sociodemographic factors and device use by cohort

	Overall (n=8,358)	Children (ages 8-12)			Adolescents (14-18)	
		ABCD (n=2,593)	Generation R (n=3,304)	INMA young (n=1,311)	HERMES (n=892)	INMA Menorca (n=258)
Sociodemographic factors						
Age (mean years (SD))	11.0 (2.1)	12.0 (0.2)	9.7 (0.3)	8.7 (0.7)	14.0 (0.9)	17.6 (0.2)
Female (vs male) (%)	50.9	50.3	50.7	49.1	56.1	52.0
Age of mother at birth (mean years, SD)	31.7 (4.5)	32.3 (4.2)	31.1 (4.8)	32.1 (3.9)	29.4 (4.1)	30.2 (4.6)
Marital status of mother (% living with partner vs living alone)	88.1	81.5	88.6	99.7	85.3	100.0
Highest level of maternal education (% university or higher)	54.6	75.3	53.5	38.6	32.6	16.4
Parity (% >2 children vs ≤ 2) ¹	13.8	8.8	11.5	6.0	49.2	8.6
Device use habits						
DECT home phone (% yes vs no)	80.2	83.4	79.8	68.8	91.7	71.6
daily median minutes (min-max)	0.4 (0-100)	0.3 (0-100)	0.3 (0-90)	0.0 (0-57)	2.5 (0-61)	0.1 (0-180)
Use of mobile phone (% yes vs no)	60.0	82.0	49.9	9.8	95.3	99.2
daily median call minutes (min-max)	0.7 (0-300)	0.7 (0-180)	0.4 (0-180)	0.0 (0-80)	6.4 (0-300)	1.4 (0-180)
Texting (% yes vs no)	63.2	77.2	53.0	31.9	95.3	99.2
daily median n of texts (min-max)	0.4 (0-1200)	1.4 (0-400)	0.1 (0-500)	0.0 (0-30)	30.5 (0-62)	72.5 (0-1200)
Browsing (% yes vs no)	59.6	76.7	54.2	23.6	75.5	83.9
daily median minutes (min-max)	2.1 (0-360)	12.9 (0-360)	0.0 (0-360)	0.0 (0-90)	53.6 (0-116)	15.0 (0-360)
Emailing (% yes vs no)	15.1	15.9	14.1	4.4	-	72.1
daily median minutes (min-max)	0.0 (0-400)	0.0 (0-120)	0.0 (0-120)	0.0 (0-60)	-	4.3 (0-400)
Use of laptop (% yes vs no)	87.5	93.0	89.1	70.8	-	65.2

connected to internet (% yes vs no) ²	90.8	95.1	91.2	81.8	-	72.1
daily median minutes (min-max)	19.1 (0-514)	24.7 (0-376)	18.9 (0-423)	8.6 (0-223)	-	30.0 (0-514)
Use of tablet (% yes vs no)	77.9	94.0	89.6	67.8	27.3	51.7
connected to internet (% yes vs no) ²	89.7	95.7	90.3	73.8	98.4	60.0
daily median minutes (min-max)	24.3 (0-635)	30.4 (0-442)	30.0 (0-635)	12.1 (0-231)	0.0 (0-350)	6.1 (0-90)

Abbreviations: ABCD, Amsterdam Born Children and their Development; DECT, digital enhanced cordless telecommunications; HERMES, Health Effects Related to Mobile phone use in adolescents; INMA, Spanish Childhood and Environment Project

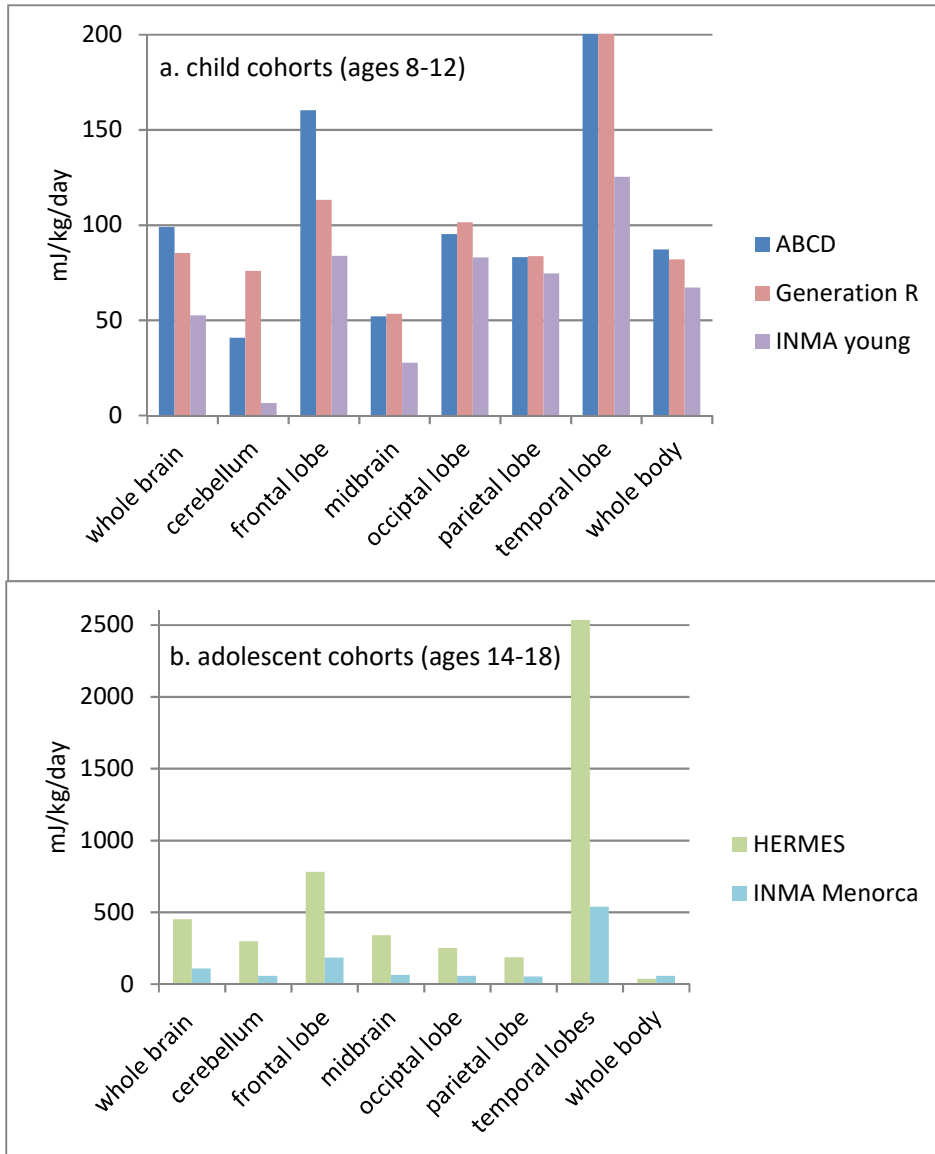
¹In HERMES, parity was not collected, children reported number of siblings instead. ²Among children that use device.

- In HERMES, emailing on mobile phone was not collected. Participants reported laptop and tablet use together, we have reported it as tablet use.

Table 3: Median modeled radiofrequency dose (mJ/kg/day) in brain regions and whole-body, overall, by age group, and by cohort

	Overall	Children	Adolescents	ABCD	Generation R	INMA young	HERMES	INMA Menorca
n	8,358	7,208	1,150	2,593	3,304	1,311	892	258
whole-brain	91.7	83.7	330.4	98.9	85.3	52.6	451.9	107.1
cerebellum	46.1	38.4	219.0	40.8	76.0	6.4	297.7	56.4
frontal lobe	136.9	123.4	582.9	160.3	113.2	83.8	781.9	183.4
midbrain	53.9	48.9	246.0	52.1	53.4	27.7	340.2	63.6
occipital lobe	101.1	96.1	180.6	95.3	101.5	83.0	249.8	57.2
parietal lobe	85.7	82.2	147.9	83.1	83.6	74.5	186.5	51.8
temporal lobes	316.9	274.9	1786.5	343.5	306.5	125.3	2535.9	537.2
whole-body	76.9	81.8	41.9	87.2	82.0	67.2	37.9	56.7

461 Abbreviations: ABCD, Amsterdam Born Children and their Development; HERMES, Health Effects Related to Mobile
 462 phonE use in adolescents; INMA, Spanish Childhood and Environment Project.



465

466

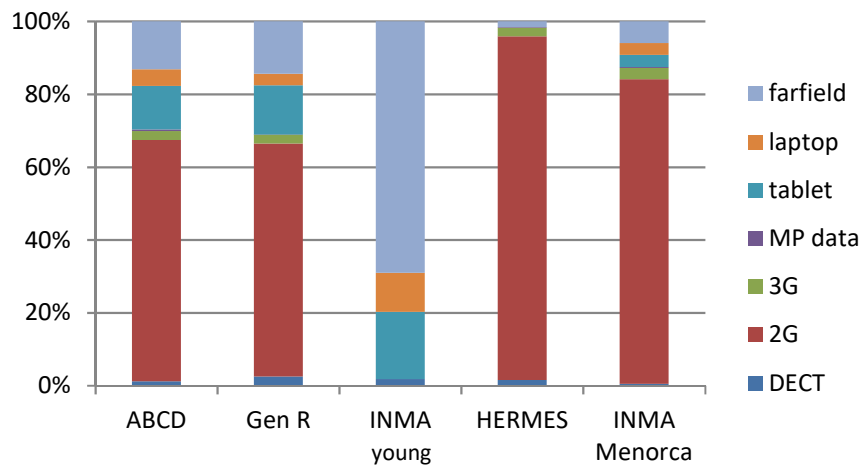
467

468

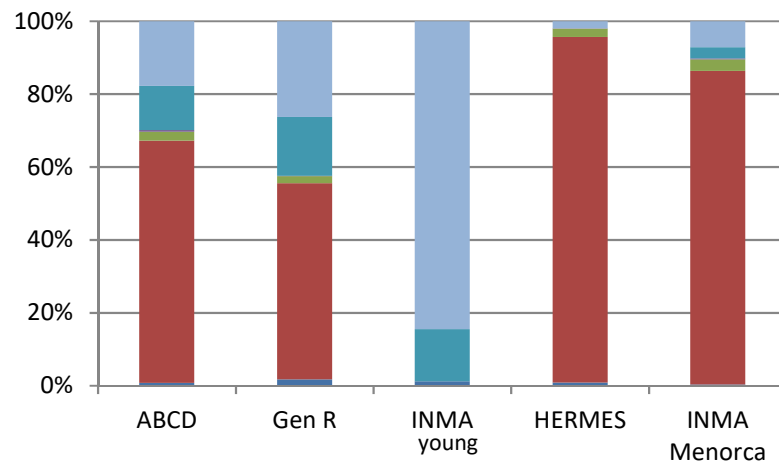
Figure 1. Median daily RF dose (mJ/kg/day) in brain regions and whole-body by cohort among **a)** child cohorts (smaller scale) and **b)** adolescent cohorts (larger scale)

469
470
471
472
473
474
475
476
477
478
479
480
481
482
483
484
485
486
487
488

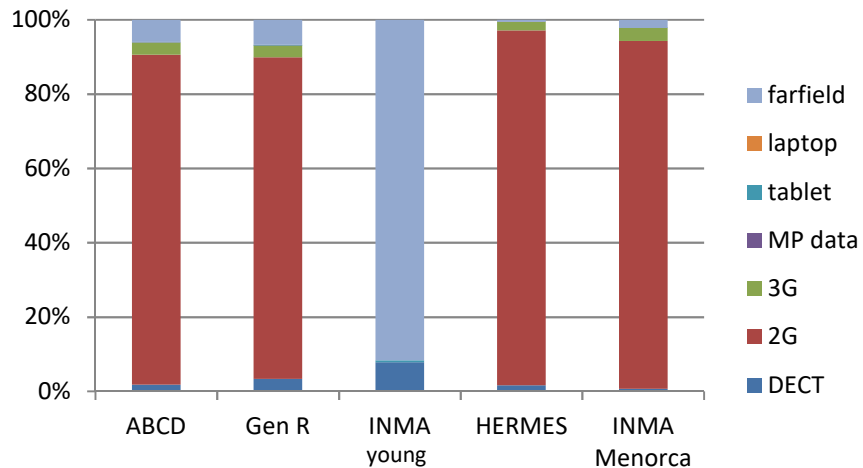
2a. whole brain



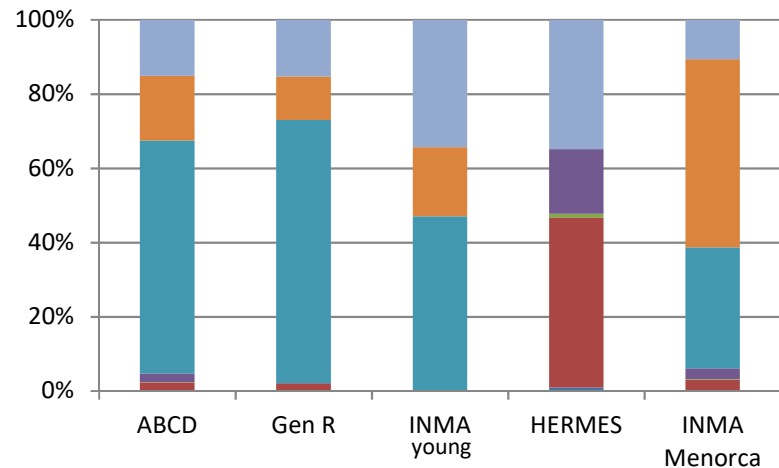
2b. frontal lobe



2c. temporal lobes



2d. whole body



489 **References**

- 490 Abramson, Michael J., Geza P. Benke, Christina Dimitriadis, Imo O. Inyang, Malcolm R. Sim, Rory S.
491 Wolfe, and Rodney J. Croft. 2009. "Mobile Telephone Use Is Associated with Changes in
492 Cognitive Function in Young Adolescents." *Bioelectromagnetics* 30 (8): 678–86.
493 <https://doi.org/10.1002/bem.20534>.
- 494 Andone, Ionut, Konrad Błaskiewicz, Mark Eibes, Boris Trendafilov, Christian Montag, and
495 Alexander Markowetz. 2016. "How Age and Gender Affect Smartphone Usage." In
496 *Proceedings of the 2016 ACM International Joint Conference on Pervasive and Ubiquitous*
497 *Computing: Adjunct*, 9–12. UbiComp '16. New York, NY, USA: ACM.
498 <https://doi.org/10.1145/2968219.2971451>.
- 499 Baars, Bernard J., and Nicole M. Gage. 2010. "Chapter 5 - The Brain." In *Cognition, Brain, and*
500 *Consciousness (Second Edition)*, edited by Bernard J. Baars and Nicole M. Gage, 126–54.
501 London: Academic Press. <https://doi.org/10.1016/B978-0-12-375070-9.00005-X>.
- 502 Beekhuizen, J., R. Vermeulen, H. Kromhout, A. Bürgi, and A. Huss. 2013. "Geospatial Modelling of
503 Electromagnetic Fields from Mobile Phone Base Stations." *Science of The Total*
504 *Environment* 445–446 (February): 202–9. <https://doi.org/10.1016/j.scitotenv.2012.12.020>.
- 505 Birks, Laura Ellen, Benjamin Struchen, Marloes Eeftens, Luuk van Wel, Anke Huss, Peter Gajšek,
506 Leeka Kheifets, et al. 2018. "Spatial and Temporal Variability of Personal Environmental
507 Exposure to Radio Frequency Electromagnetic Fields in Children in Europe." *Environment*
508 *International* 117 (August): 204–14. <https://doi.org/10.1016/j.envint.2018.04.026>.
- 509 Bürgi, Alfred, Patrizia Frei, Gaston Theis, Evelyn Mohler, Charlotte Braun-Fahrländer, Jürg Fröhlich,
510 Georg Neubauer, Matthias Egger, and Martin Röösli. 2010. "A Model for Radiofrequency
511 Electromagnetic Field Predictions at Outdoor and Indoor Locations in the Context of
512 Epidemiological Research." *Bioelectromagnetics* 31 (3): 226–36.
513 <https://doi.org/10.1002/bem.20552>.
- 514 Cabré-Riera, Alba, Hanan El Marroun, Ryan Muetzel, Luuk van Wel, Ilaria Liorni, Arno Thielens,
515 Laura Ellen Birks, et al. 2020. "Estimated Whole-Brain and Lobe-Specific Radiofrequency
516 Electromagnetic Fields Doses and Brain Volumes in Preadolescents." *Environment*
517 *International* 142 (September): 105808. <https://doi.org/10.1016/j.envint.2020.105808>.
- 518 Cabré-Riera, Alba, Luuk vanWel, Ilaria Liorni, Arno Thielens, Laura Ellen Birks, Livia Pierotti, Wout
519 Joseph, et al. 2020. "Association between Estimated Whole-Brain Radiofrequency
520 Electromagnetic Fields Dose and Cognitive Function in Preadolescents and Adolescents."
521 *International Journal of Hygiene and Environmental Health* In Press.
- 522 Calvente, I., M. F. Fernández, R. Pérez-Lobato, C. Dávila-Arias, O. Ocón, R. Ramos, S. Ríos-Arrabal, J.
523 Villalba-Moreno, N. Olea, and M. I. Núñez. 2015. "Outdoor Characterization of Radio
524 Frequency Electromagnetic Fields in a Spanish Birth Cohort." *Environmental Research* 138
525 (April): 136–43. <https://doi.org/10.1016/j.envres.2014.12.013>.

- 526 Eeftens, Marloes, Benjamin Struchen, Laura Ellen Birks, Elisabeth Cardis, Marisa Estarlich, Mariana
527 F. Fernandez, Peter Gajšek, et al. 2018. "Personal Exposure to Radio-Frequency
528 Electromagnetic Fields in Europe: Is There a Generation Gap?" *Environment International*
529 121 (December): 216–26. <https://doi.org/10.1016/j.envint.2018.09.002>.
- 530 Eijdsden, Manon van, Tanja GM Vrijkotte, Reinoud JBJ Gemke, and Marcel F. van der Wal. 2011.
531 "Cohort Profile: The Amsterdam Born Children and Their Development (ABCD) Study."
532 *International Journal of Epidemiology* 40 (5): 1176–86.
533 <https://doi.org/10.1093/ije/dyq128>.
- 534 Feychting, Maria. 2011. "Mobile Phones, Radiofrequency Fields, and Health Effects in Children--
535 Epidemiological Studies." *Progress in Biophysics and Molecular Biology* 107 (3): 343–48.
536 <https://doi.org/10.1016/j.pbiomolbio.2011.09.016>.
- 537 Foerster, Milena, Arno Thielens, Wout Joseph, Marloes Eeftens, and Martin Röösli. 2018. "A
538 Prospective Cohort Study of Adolescents' Memory Performance and Individual Brain Dose
539 of Microwave Radiation from Wireless Communication." *Environmental Health*
540 *Perspectives* 126 (7): 077007. <https://doi.org/10.1289/EHP2427>.
- 541 Goedhart, Geertje, Luuk van Wel, Chelsea E. Langer, Patricia de Llobet Viladoms, Joe Wiart,
542 Martine Hours, Hans Kromhout, et al. 2018. "Recall of Mobile Phone Usage and Laterality
543 in Young People: The Multinational Mobi-Expo Study." *Environmental Research* 165
544 (August): 150–57. <https://doi.org/10.1016/j.envres.2018.04.018>.
- 545 Gosselin, Marie-Christine, Esra Neufeld, Heidi Moser, Eveline Huber, Silvia Farcito, Livia Gerber,
546 Maria Jedensjö, et al. 2014. "Development of a New Generation of High-Resolution
547 Anatomical Models for Medical Device Evaluation: The Virtual Population 3.0." *Physics in*
548 *Medicine & Biology* 59 (18): 5287. <https://doi.org/10.1088/0031-9155/59/18/5287>.
- 549 Graham, John W., Allison E. Olchowski, and Tamika D. Gilreath. 2007. "How Many Imputations Are
550 Really Needed? Some Practical Clarifications of Multiple Imputation Theory." *Prevention*
551 *Science* 8 (3): 206–13. <https://doi.org/10.1007/s11121-007-0070-9>.
- 552 Guxens, Mònica, Ferran Ballester, Mercedes Espada, Mariana F. Fernández, Joan O. Grimalt, Jesús
553 Ibarluzea, Nicolás Olea, et al. 2012. "Cohort Profile: The INMA—INfancia y Medio
554 Ambiente—(Environment and Childhood) Project." *International Journal of Epidemiology*
555 41 (4): 930–40. <https://doi.org/10.1093/ije/dyr054>.
- 556 Guxens, Mònica, Roel Vermeulen, Manon van Eijdsden, Johan Beekhuizen, Tanja G. M. Vrijkotte,
557 Rob T. van Strien, Hans Kromhout, and Anke Huss. 2016. "Outdoor and Indoor Sources of
558 Residential Radiofrequency Electromagnetic Fields, Personal Cell Phone and Cordless
559 Phone Use, and Cognitive Function in 5–6 Years Old Children." *Environmental Research*
560 150 (October): 364–74. <https://doi.org/10.1016/j.envres.2016.06.021>.
- 561 Huss, Anke, Manon van Eijdsden, Monica Guxens, Johan Beekhuizen, Rob van Strien, Hans
562 Kromhout, Tania Vrijkotte, and Roel Vermeulen. 2015. "Environmental Radiofrequency
563 Electromagnetic Fields Exposure at Home, Mobile and Cordless Phone Use, and Sleep

564 Problems in 7-Year-Old Children." *PLOS ONE* 10 (10): e0139869.
565 <https://doi.org/10.1371/journal.pone.0139869>.

566 Joseph, Wout, Daan Pareit, Günter Vermeeren, Dries Naudts, Leen Verloock, Luc Martens, and
567 Ingrid Moerman. 2013. "Determination of the Duty Cycle of WLAN for Realistic Radio
568 Frequency Electromagnetic Field Exposure Assessment." *Progress in Biophysics and
569 Molecular Biology* 111 (1): 30–36. <https://doi.org/10.1016/j.pbiomolbio.2012.10.002>.

570 Kheifets, Leeka, Michael Repacholi, Rick Saunders, and Emilie van Deventer. 2005. "The Sensitivity
571 of Children to Electromagnetic Fields." *Pediatrics* 116 (2): e303–13.
572 <https://doi.org/10.1542/peds.2004-2541>.

573 Kooijman, Marjolein N., Claudia J. Kruithof, Cornelia M. van Duijn, Liesbeth Duijts, Oscar H. Franco,
574 Marinus H. van IJzendoorn, Johan C. de Jongste, et al. 2016. "The Generation R Study:
575 Design and Cohort Update 2017." *European Journal of Epidemiology* 31 (12): 1243–64.
576 <https://doi.org/10.1007/s10654-016-0224-9>.

577 Langer, Chelsea E., Patricia de Llobet, Albert Dalmau, Joe Wiart, Geertje Goedhart, Martine Hours,
578 Geza P. Benke, et al. 2017. "Patterns of Cellular Phone Use among Young People in 12
579 Countries: Implications for RF Exposure." *Environment International* 107 (October): 65–74.
580 <https://doi.org/10.1016/j.envint.2017.06.002>.

581 Lauer, Oliver, Patrizia Frei, Marie-Christine Gosselin, Wout Joseph, Martin Rösli, and Jürg
582 Fröhlich. 2013. "Combining Near- and Far-Field Exposure for an Organ-Specific and Whole-
583 Body RF-EMF Proxy for Epidemiological Research: A Reference Case." *Bioelectromagnetics*
584 34 (5): 366–74. <https://doi.org/10.1002/bem.21782>.

585 Liorni, Ilaria, Myles Capstick, Luuk van Wel, Joe Wiart, Wout Joseph, Elisabeth Cardis, Mònica
586 Guxens, Roel Vermeulen, and Arno Thielens. 2020. "EVALUATION OF SPECIFIC
587 ABSORPTION RATE IN THE FAR-FIELD, NEAR-TO-FAR FIELD AND NEAR-FIELD REGIONS FOR
588 INTEGRATIVE RADIOFREQUENCY EXPOSURE ASSESSMENT." *Radiation Protection
589 Dosimetry* 190 (4): 459–72. <https://doi.org/10.1093/rpd/ncaa127>.

590 Markov, M., and Y. Grigoriev. 2015. "Protect Children from EMF." *Electromagnetic Biology and
591 Medicine* 34 (3): 251–56. <https://doi.org/10.3109/15368378.2015.1077339>.

592 Otto, Matthias, and Karl Ernst von Mühlendahl. 2007. "Electromagnetic Fields (EMF): Do They Play
593 a Role in Children's Environmental Health (CEH)?" *International Journal of Hygiene and
594 Environmental Health* 210 (5): 635–44. <https://doi.org/10.1016/j.ijheh.2007.07.007>.

595 Persson, Tomas, Christer Törnevik, Lars-Eric Larsson, and Jan Lovén. 2012. "Output Power
596 Distributions of Terminals in a 3G Mobile Communication Network." *Bioelectromagnetics*
597 33 (4): 320–25. <https://doi.org/10.1002/bem.20710>.

598 R Core Team. 2013. *R: The R Project for Statistical Computing*. R Foundation for Statistical
599 Computing, Vienna, Austria. <https://www.r-project.org/>.

- 600 Rice, D., and S. Barone. 2000. "Critical Periods of Vulnerability for the Developing Nervous System:
601 Evidence from Humans and Animal Models." *Environmental Health Perspectives* 108 Suppl
602 3 (June): 511–33.
- 603 Rosenberg, Suzanne. 2013. "Cell Phones and Children: Follow the Precautionary Road." *Pediatric*
604 *Nursing* 39 (2): 65–70.
- 605 Roser, Katharina, Anna Schoeni, Alfred Bürgi, and Martin Rössli. 2015. "Development of an RF-
606 EMF Exposure Surrogate for Epidemiologic Research." *International Journal of*
607 *Environmental Research and Public Health* 12 (5): 5634–56.
608 <https://doi.org/10.3390/ijerph120505634>.
- 609 Roser, Katharina, Anna Schoeni, Benjamin Struchen, Marco Zahner, Marloes Eeftens, Jürg Fröhlich,
610 and Martin Rössli. 2017. "Personal Radiofrequency Electromagnetic Field Exposure
611 Measurements in Swiss Adolescents." *Environment International* 99 (February): 303–14.
612 <https://doi.org/10.1016/j.envint.2016.12.008>.
- 613 Rosso, Isabelle M., Ashley D. Young, Lisa A. Femia, and Deborah A. Yurgelun-Todd. 2004.
614 "Cognitive and Emotional Components of Frontal Lobe Functioning in Childhood and
615 Adolescence." *Annals of the New York Academy of Sciences* 1021 (June): 355–62.
616 <https://doi.org/10.1196/annals.1308.045>.
- 617 Schoeni, Anna, Katharina Roser, Alfred Bürgi, and Martin Rössli. 2016. "Symptoms in Swiss
618 Adolescents in Relation to Exposure from Fixed Site Transmitters: A Prospective Cohort
619 Study." *Environmental Health* 15: 77. <https://doi.org/10.1186/s12940-016-0158-4>.
- 620 Schoeni, Anna, Katharina Roser, and Martin Rössli. 2015. "Memory Performance, Wireless
621 Communication and Exposure to Radiofrequency Electromagnetic Fields: A Prospective
622 Cohort Study in Adolescents." *Environment International* 85 (December): 343–51.
623 <https://doi.org/10.1016/j.envint.2015.09.025>.
- 624 Schoeni, Anna, Katharina Roser, and Martin Rössli. 2017. "Symptoms and the Use of Wireless
625 Communication Devices: A Prospective Cohort Study in Swiss Adolescents." *Environmental*
626 *Research* 154: 275–83. <https://doi.org/10.1016/j.envres.2017.01.004>.
- 627 Schüz, Joachim. 2005. "Mobile Phone Use and Exposures in Children." *Bioelectromagnetics* 26 (S7):
628 S45–50. <https://doi.org/10.1002/bem.20129>.
- 629 Sterne, Jonathan A. C., Ian R. White, John B. Carlin, Michael Spratt, Patrick Royston, Michael G.
630 Kenward, Angela M. Wood, and James R. Carpenter. 2009. "Multiple Imputation for
631 Missing Data in Epidemiological and Clinical Research: Potential and Pitfalls." *BMJ* 338
632 (June): b2393. <https://doi.org/10.1136/bmj.b2393>.
- 633 Sudan, Madhuri, Jorn Olsen, Oyebuchi A. Arah, Carsten Obel, and Leeka Kheifets. 2016.
634 "Prospective Cohort Analysis of Cellphone Use and Emotional and Behavioural Difficulties
635 in Children." *Journal of Epidemiology and Community Health*, May, jech-2016-207419.
636 <https://doi.org/10.1136/jech-2016-207419>.
- 637 Szklo, M. 1998. "Population-Based Cohort Studies." *Epidemiologic Reviews* 20 (1): 81–90.

- 638 Thomas, Silke, Sabine Heinrich, Rüdiger von Kries, and Katja Radon. 2009. "Exposure to Radio-
639 Frequency Electromagnetic Fields and Behavioural Problems in Bavarian Children and
640 Adolescents." *European Journal of Epidemiology* 25 (2): 135–41.
641 <https://doi.org/10.1007/s10654-009-9408-x>.
- 642 Thomas, Silke, Sabine Heinrich, Anja Kühnlein, and Katja Radon. 2010. "The Association between
643 Socioeconomic Status and Exposure to Mobile Telecommunication Networks in Children
644 and Adolescents." *Bioelectromagnetics* 31 (1): 20–27. <https://doi.org/10.1002/bem.20522>.
- 645 Thomas, Silke, Anja Kühnlein, Sabine Heinrich, Georg Praml, Rüdiger von Kries, and Katja Radon.
646 2008. "Exposure to Mobile Telecommunication Networks Assessed Using Personal
647 Dosimetry and Well-Being in Children and Adolescents: The German MobilEe-Study."
648 *Environmental Health* 7: 54. <https://doi.org/10.1186/1476-069X-7-54>.
- 649 vanWel, Luuk, Ilaria Liorni, Anke Huss, Arno Thielens, Joe Wiart, Wout Joseph, Myles Capstick,
650 Elisabeth Cardis, and Roel Vermeulen. 2020. "Radio-Frequency Electromagnetic Field
651 Exposure and Contribution of Sources in the General Population: An Organ-Specific
652 Integrative Exposure Assessment." *Journal of Exposure Science & Environmental
653 Epidemiology* In press.
- 654