

1 **Evaluation of the CALIOPE air quality forecasting system for epidemiological**
2 **research: the example of NO₂ in the province of Girona (Spain)**

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48 **Abstract**

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50 Background: Air quality models are being increasingly used to estimate long-term
51 individual exposures to air pollution in epidemiological studies. Most of them have been
52 evaluated against measurements from a limited number of monitoring stations, which
53 may not properly reflect the exposure characteristics of the study population.

54 Methods: We evaluated the performance of the high-resolution CALIOPE air quality
55 forecasting system over a large sample of passive measurements of NO₂ conducted at
56 635 home outdoor locations of the Girona province (Spain) during several 4-week
57 sampling campaigns over one year. Sampling sites were superposed over the 4 km x 4
58 km CALIOPE grid, and average NO₂ modeled concentrations were derived for all
59 measurements conducted during the same sampling campaign at all the sampling sites
60 located within the same grid cell. In addition, the ratio between measured and modeled
61 concentrations for the whole study period at one fixed monitoring station was used to
62 post-process the modeled values at the home outdoor locations.

63 Results: The correlation between measured and modeled concentrations for the entire
64 study area (which includes urban settings, middle-size towns, and rural areas) was 0.78.
65 After correcting the modeled concentrations by the measured to modeled ratio at the
66 fixed station, they were very similar to the measured concentrations (27.7 µg/m³ and
67 29.3 µg/m³, respectively). However, the performance of the modeling system depends
68 on the type of subarea and is affected by the sub-grid emission sources.

69 Conclusions: The evaluation over the heterogeneous Girona province showed that
70 CALIOPE is able to reproduce the spatial variability of 4-week NO₂ concentrations at
71 the small regional level, but not at the smaller within-city scale. CALIOPE output data
72 is as a valuable tool to complement study-specific air pollution measurements by
73 incorporating regional spatial variability and long-term temporal variability of
74 background pollution in epidemiological research.

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76 **Keywords:** air quality modeling; exposure assessment; model evaluation; NO₂

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83 **1. Introduction**

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85 Air pollution is the most important environmental factor affecting human health in
86 Europe, contributing in a non-negligible manner to the total burden of mortality in
87 urban areas (Boldo et al., 2006; Künzli et al, 2000). As a result, air quality management
88 is a priority issue in European environmental policies. Modeling techniques have
89 become very useful tools to study the dynamics and transport of air pollutants and to
90 forecast air quality concentration for short term mitigation and public information and
91 warnings, both applications for regulatory and scientific purposes (European
92 Commission, 2008; EEA, 2011).

93 From the epidemiological research perspective, there is an increasing interest in using
94 data from air quality models to estimate long-term individual exposures to air pollution
95 in population-based studies. This interest lies in the ability of air quality models to
96 reflect the spatial variability of air pollution both in the short- and long term, in contrast
97 with other techniques such as land use regression (LUR) models, which usually better
98 capture the spatial variability of air pollution at local scales but with a much lower
99 temporal resolution (Marshall et al., 2008). In most cases, air quality models have not
100 been evaluated in the particular study area or have been evaluated against measurements
101 from a limited number of fixed monitoring stations (Bellander et al., 2001; Cyrus et al.,
102 2005; Nyberg et al., 2000; Van den Hooven et al. 2012). Although air quality networks
103 have the advantage of providing a good temporal coverage with routinely available
104 regulatory air pollution data, data from a few stations provide poor spatial resolution to
105 validate air quality models at the small regional or local scales. In addition, air quality
106 networks are typically designed to assess air pollution levels for policy and regulatory
107 purposes, not for health studies. Therefore, they may not properly reflect the exposure
108 characteristics of the population living in a given study area, particularly for those
109 pollutants that are more spatially heterogeneous (Brauer, 2010). For this reason, output
110 data from air quality models should be preferably validated against observations in
111 targeted locations selected to represent the real range of individual-level exposure
112 within a study population. To our knowledge, only one study has been able to evaluate
113 air pollution concentrations predicted by a dispersion model against NO₂ observations
114 measured with passive samplers in a sample of residential outdoor locations from a
115 Swiss population-based epidemiological study (Liu et al., 2007).

116 The present study aims to evaluate the CALIOPE air quality forecasting system as a
117 tool for incorporating regional spatial variability and long-term temporal variability in
118 air pollution exposure estimates for the participants of an epidemiological study
119 (REGICOR) in the Girona province (northeast Spain). The CALIOPE project, funded
120 by the Spanish Ministry of the Environment, has established a high-resolution air
121 quality forecasting system for Spain (Baldasano et al., 2008b). The CALIOPE system
122 combines four models to simulate air quality over Spain with a high spatial resolution (4
123 km x 4 km in the Iberian Peninsula domain), and a temporal resolution of 1 hour using
124 the HERMES emission model specifically built up for Spain (Baldasano et al., 2008a).
125 So far, CALIOPE is the only European system which includes the Saharan dust
126 contribution on an hourly basis. Several evaluation studies (Baldasano et al., 2011; Pay
127 et al., 2010; Pay et al., 2012) and near-real time evaluation (NRT) against air quality
128 measurements on an hourly basis support the confidence on the system. Current air
129 quality forecasts and NRT evaluations are available through <http://www.bsc.es/caliope>.
130 The performance of the CALIOPE modeling system has been evaluated by comparing
131 its estimates with measured concentrations of particulate matter less than 10 μm in
132 aerodynamic diameter (PM_{10}) and gaseous pollutants (NO_2 , SO_2 , and O_3) at the Spanish
133 network of air quality monitoring stations for a full year (Baldasano et al., 2011). Strong
134 differences were found depending on the type of site (urban, suburban, and rural) and
135 the main emission sources (traffic, industrial, and background). For NO_2 , modeled
136 concentrations showed good agreement with observations, particularly in rural areas
137 influenced by background emissions, although levels were systematically
138 underestimated in all background and many urban sites. A similar pattern was found for
139 PM_{10} measurements. On the contrary, modeled O_3 levels performed better in urban
140 settings, especially at traffic stations located in the largest cities (Madrid and
141 Barcelona).

142 Given the importance of assessing the performance of air quality models in study-
143 specific targeted locations for epidemiological research, the REGICOR study, which
144 has conducted a large amount of NO_2 measurements at residential outdoor locations
145 specifically selected to cover a broad range of traffic-related air pollution and urban
146 settings, offers a unique opportunity to evaluate the performance of the CALIOPE
147 modeling system for estimating NO_2 concentrations at a geographically-diverse
148 province level.

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150 **2. Material and methods**

151

152 *2.1. CALIOPE modeling system description*

153 The state-of-the-art CALIOPE modeling system has been described in detail elsewhere
154 (Baldasano et al., 2008b; 2011). Briefly, it integrates a meteorological model
155 (WRFv3.0.1.1), an emission model (HERMES 2004), a chemical transport model
156 (CMAQv4.5), and a mineral dust atmospheric model (BSC-DREAM8b) together in an
157 air quality modeling and forecasting system.

158 The Advanced Research Weather Research and Forecasting model (WRF-ARW)
159 (Skamarock and Klemp, 2008) provides the meteorological fields required by the
160 chemical transport model. The model top is defined at 50 hPa to resolve the
161 troposphere-stratosphere exchanges properly. The High-Selective Resolution Modeling
162 Emission System (HERMES) (Baldasano et al., 2008a) uses information and the state-
163 of-the-art methodologies for emission estimations. It calculates emissions by sector-
164 specific sources or by individual installation and stacks following a bottom-up approach
165 for the reference year 2004. The Models-3 Community Multiscale Air Quality
166 Modeling System (Models-3/CMAQ) (Byun and Schere, 2006) combines current
167 knowledge in atmospheric science and air quality modeling with multi-processor
168 computing techniques in an open-source framework to deliver concentration estimates
169 of the main air pollutants (O₃, NO₂, SO₂, and PM₁₀, among others) and acid deposition.
170 Finally, CALIOPE modeled particulate matter are achieved by adding the Saharan dust
171 contribution from BSC-DREAM8bin (Pérez et al., 2006a; 2006b) model to the CMAQ
172 outputs.

173 Hourly NO₂ concentrations from the CALIOPE system over one-year period (from 1st
174 July 2007 to 31st June 2008) in a 4 km x 4 km grid covering the province of Girona
175 were used for the present study.

176

177 *2.2. NO₂ measurements in REGICOR*

178 REGICOR (Girona Heart Registry) is a population-based study which aims at
179 investigating the effect of chronic exposure to ambient air pollution on subclinical
180 markers of atherosclerosis in the Girona province. Twelve of the original towns
181 included in REGICOR that provided the maximum contrast in ambient air pollution
182 levels in a pilot study, while keeping the travel distance between towns to a minimum,
183 were selected for the present study. The final study area has approx. 4,550 km² and

184 includes the average mid-size city of Girona and adjacent Salt (96,200 and 30,300
185 inhabitants, respectively), towns on the Mediterranean coast with maritime ports and
186 tourist resorts, inland towns, and small villages located in valleys and mountains (Figure
187 1). In countries such as Spain, fixed site monitoring stations are rare in regions with
188 small communities clustered in otherwise rather rural areas. The present study area has
189 only one monitoring station, located at an urban background site in Girona city.
190 From the total number of 2,780 subjects who underwent a detailed cardiovascular health
191 examination, we selected a subsample of 635 subjects for home outdoor NO₂
192 monitoring with passive Palmes tubes over 4 weeks (Palmes et al., 1976) sent to the
193 participants. Homes were selected to cover a broad range of traffic characteristics and
194 urban settings (e.g. low and high building-density areas, historical quarters, and
195 peripheral residential areas), to be representative of the residential locations of the
196 REGICOR participants, and to be well distributed across the 12 towns. In addition, 4-
197 week NO₂ monitoring was performed every month at a reference site in each town for
198 an entire year, in order to derive temporally-adjusted NO₂ concentrations in the
199 sampling sites of the corresponding town. Results from our pilot study and from
200 previous monitoring studies conducted by local authorities indicated that monthly mean
201 NO₂ concentrations during spring and fall were similar to the annual mean. Thus, from
202 the 23 sampling campaigns that were performed from June 2007 to July 2009, the
203 largest ones were conducted during spring and fall. For the present study, NO₂
204 measurements performed between 1st July 2007 and 31st June 2008 were included.

205

206 *2.3. CALIOPE NO₂ estimates for REGICOR sampling sites*

207 Home addresses of REGICOR participants were geocoded at the front door level. Then,
208 the geocoded addresses were superposed over the CALIOPE grid to identify the
209 corresponding 4 km x 4 km cell for each address (Figure 2), and the average NO₂
210 modeled concentration was derived for each sampling site (home address) and for the
211 corresponding 4-weeks sampling period, matching the period from the opening until the
212 closing day of each passive sampler. In addition, daily NO₂ modeled concentrations
213 were derived for the urban background station located in Girona city over the whole
214 study period.

215

216 *2.4. Statistical analysis*

217 A first comparison was made between measured and modeled NO₂ concentrations at the
218 fixed monitoring station in Girona city over the whole study period. Since the sampling
219 period of Palmes tubes was 4 weeks, for each day of the study period we computed a 4-
220 week average of measured and modeled concentrations, being the reference day the start
221 date of the 4-weeks time window. The ratio between 4-weeks measured and modeled
222 concentrations was derived for each day, and the average ratio for the whole study
223 period was used to post-process the modeled values at the home outdoor sampling sites.
224 In order to compare modeled NO₂ concentrations with the passive tubes measurements,
225 we averaged all measurements conducted during the same sampling campaign at all the
226 sampling sites located within the same 4 km x 4 km cell. Because the number of
227 measurements taken at each sampling site over the study period was heterogeneous, and
228 the number of sampling sites located within a same grid cell differed across the study
229 area, we additionally compared modeled and measured concentrations weighted by the
230 total number of measurements used to obtain the average value at each grid cell for the
231 same sampling campaign.

232 Descriptive statistics of both measured and modeled concentrations were calculated for
233 the whole study area and for six different geographic clusters of adjacent towns, which
234 reflect similarities in urban characteristics, topography, and traffic intensity levels.
235 Correlation coefficients and root mean square errors (RMSE) were calculated to assess
236 the performance of the CALIOPE modeling system over the sampling sites at the
237 different groups of towns.

238 Analyses were performed using Stata 10.1 (StataCorp, College Station, TX) and R 2.12
239 (<http://www.R-project.org>).

240

241 **3. Results**

242

243 The correlation between daily measured and modeled NO₂ concentrations at the urban
244 background monitoring station of Girona city was moderate for the whole study period
245 (Pearson's $r=0.35$) but low for the extreme seasons, i.e. winter and summer (Pearson's
246 $r<0.10$). Modeled NO₂ concentrations were persistently underestimated (mean=8.8
247 $\mu\text{g}/\text{m}^3$, SD=3.3 $\mu\text{g}/\text{m}^3$) as compared to measured concentrations (mean=35.7 $\mu\text{g}/\text{m}^3$,
248 SD=10.4 $\mu\text{g}/\text{m}^3$). The average ratio of 4-week modeled to measured concentrations for
249 the whole study period at the fixed monitoring station was 0.25. This ratio was used to
250 post-process the modeled NO₂ concentrations in all the sampling sites of the study area.

251 A scatter plot of 4-week measured and modeled (corrected by the average ratio) NO₂
252 concentrations at the fixed monitoring station of Girona is shown in Figure 3. The
253 highest modeled concentrations (>70 µg/m³), mainly corresponding to the winter
254 season, are overestimated by the model after post-processing, which contributes to
255 reduce the correlation found for this period.

256 Table 1 shows the descriptive statistics for measured and modeled 4-week NO₂
257 concentrations in the whole REGICOR area and by groups of towns. After averaging
258 the number of passive measurements conducted during the same sampling campaign
259 and located within the same grid cell, a total number of 165 measurements were used to
260 assess the performance of the CALIOPE modeling system. The adjusted mean modeled
261 NO₂ concentration for the whole study area was very similar to the mean measured
262 concentration (27.7 and 29.3 µg/m³, respectively). The Spearman correlation between
263 modeled and measured concentrations for the entire study area was high (r=0.78). When
264 the six groups of towns were considered independently, correlations were moderate to
265 high (r=0.40 – 0.76) and did not follow a clear geographic pattern. On the contrary, an
266 urban-rural pattern was shown by the RMSE, with higher values in the urban area of
267 Girona and Salt (11.3 µg/m³), lower values in rural areas (7.7 µg/m³ in Banyoles-
268 Porqueres and 6.7 µg/m³ in Llagostera-Santa Cristina) and intermediate values in the
269 remaining middle-size inland and coastal towns.

270 The linear relationship between modeled and measured NO₂ concentrations can be
271 visualized in Figure 4. The scatter plots show the difference in terms of absolute
272 concentrations between the urban area of Girona-Salt and the rest of the REGICOR
273 area, as well as the lack of clear outliers, neither in Girona-Salt nor in the rest of the
274 study area. When the weight of the number of tubes for each measurement site is
275 incorporated into the scatter plot (Figure 4B), the higher contribution of the reference
276 sites and the larger concentration of sampling sites over the 4 km x 4 km cells in
277 Girona-Salt are clearly highlighted, and the R² decreased from 0.61 to 0.33.

278

279 **4. Discussion**

280

281 In the present study, we have found a good correlation between NO₂ concentrations
282 measured with Palmes tubes at a large sample of residential locations and modeled NO₂
283 concentrations with the CALIOPE modeling system over one year across the
284 geographically-diverse province of Girona, Spain. Despite the high overall correlation,

285 modeled concentrations were systematically underestimated in the whole study area,
286 which includes urban settings, inland and coastal towns, and rural areas. CALIOPE
287 predictions were successfully corrected for this underestimation using continuous data
288 over the study period from the fixed monitoring station representative of the urban
289 background concentrations in the city of Girona. Modeled concentrations showed a
290 substantially higher correlation with measurements at residential locations than with
291 measurements at the fixed monitoring station. These results indicate that relying on
292 daily measurements from a small sample of monitoring sites is insufficient to properly
293 assess the performance of an air quality modeling system such as CALIOPE at the
294 province (i.e. small regional) level over long time periods.

295 The performance of the CALIOPE modeling system has been previously evaluated by
296 using hourly air pollution data from the 68 monitoring stations of the Spanish air quality
297 network over the year 2004. The correlation coefficient between measured and modeled
298 NO₂ concentrations was 0.53 for the whole sample of monitoring stations and decreased
299 when the authors split the sample into urban stations ($r=0.47$), suburban stations
300 ($r=0.39$), and rural stations ($r=0.51$) (Baldasano et al., 2011). Although the differences
301 between this previous study and the present one in terms of study area (Spain vs. Girona
302 province) and type of NO₂ measurements (automatic monitoring of hourly
303 concentrations vs. passive monitoring of monthly concentrations) make direct
304 comparisons difficult, the better performance of CALIOPE found in our study supports
305 the argument that, for epidemiological studies on air pollution and health, a
306 comprehensive evaluation of output data from air quality models should be made
307 against measurements in locations specifically selected to represent the exposure
308 distribution among the study population.

309 In most of the epidemiological studies, however, model evaluation has been often
310 limited to a few monitoring sites, usually those from the air quality networks (Bellander
311 et al., 2001; Cyrus et al., 2005; Nyberg et al., 2000; Van den Hooven et al., 2012),
312 which are unable to identify potential differences in model performance across different
313 subareas. The REGICOR study provided a unique opportunity to identify such
314 geographical variations in performance, given the intensive NO₂ passive monitoring
315 over one year at a large number of sampling sites, specifically selected to cover a broad
316 range of geographic characteristics, to reflect the highest contrast in air pollution levels,
317 and to represent the spatial distribution of the study participants. The comparison of
318 CALIOPE estimates with NO₂ measurements revealed a good performance at the small

319 regional level of a 4 km x 4 km grid but the inability to reflect the spatial gradients at
320 the smaller locale scale.

321 The underestimation of modeled NO₂ concentrations at the fixed monitoring station is
322 not surprising, since urban stations are more likely to be influenced by very local
323 emission sources which are more difficult to be captured by the model. Regarding the
324 low correlation between measured and modeled concentrations at this station, the
325 previous assessment of the CALIOPE modeling system over the whole Spanish air
326 quality monitoring network already identified that measurements at urban stations from
327 small and medium-sized cities generally showed poorer agreement with modeled
328 estimations, whereas modeled concentrations were more accurate at urban stations
329 located in larger and better characterized cities such as Barcelona or Madrid, for which
330 a higher spatial resolution (1x1 km) was available (Baldasano et al., 2011).

331 When the performance of CALIOPE was assessed against the measurements of Palmes
332 tubes across the REGICOR study area, results showed that not only urban but also
333 suburban and rural areas were systematically underestimated. The methodology used to
334 estimate traffic emissions at medium- and small-sized cities may explain this result.
335 Besides, biomass burning and natural NO_x production such as lightning are not
336 currently treated in the model, and biogenic emissions in HERMES may need further
337 revision. To correct this underestimation in the CALIOPE forecasting system, an
338 updated version of the inventory is being developed (Guevara et al., 2012).

339 Uncertainties in the simulated air pollution concentrations are common in modeling
340 techniques. Misrepresentation of atmospheric dynamics and chemistry, geographic
341 differences in the quality of input data, and the choice of modeling domain and grid
342 structure are among the causes associated with model uncertainties (Borrego et al.,
343 2008; Sicardi et al., 2012). Several post-processing techniques have been applied to
344 correct the modeled concentrations based on observed data, such as Bayesian-based
345 techniques, data assimilation, or the Kalman filter technique (Napelenok et al., 2008).

346 The latter has been successfully applied within CALIOPE to improve the prediction of
347 tropospheric ozone (Sicardi et al., 2012), and further works in Portugal have extended
348 the method to other pollutants including NO₂ (Borrego et al., 2011).

349 Although the correlation between modeled and measured concentrations is relatively
350 high in the cities of Girona and Salt, the highest mean RMSE value for this urban area is
351 indicative of a poorer performance of the model in comparison with suburban and rural
352 areas. This can be explained by two factors. First, the HERMES emission model

353 incorporates emissions data from on-road transport, but in cities with less than 500,000
354 inhabitants the availability of traffic emissions is restricted to major roads only
355 (Baldasano et al., 2008a). Thus, the influence of high and very local on-road traffic is
356 not well characterized in middle-size cities. Second, the large number of measurements
357 with Palmes tubes taken in Girona and Salt provides a wide range of NO₂
358 concentrations which are considerably diluted when the sampling sites are superposed
359 on a grid of 4 km x 4 km to obtain the predicted values, given the small area of these
360 two cities (45.7 km²).

361 The problem of comparing grid model predictions with point measurements is well
362 known, namely the within-grid cell variability in emission sources due to different land
363 uses, topography, traffic activities, and other characteristics that typically vary at finer
364 scales. This problem is partially solved, but never eliminated entirely, by using finer
365 scale grid sizes or by applying within-grid model treatments for the major point sources
366 (Ching et al., 2006). This is particularly relevant for NO₂, which usually shows a
367 heterogeneous spatial distribution at the intraurban level, with relatively large contrast
368 within short distances (Lewné et al., 2004). In fact, a study conducted in several urban
369 areas of Switzerland demonstrated that even a dispersion model with a fine spatial
370 resolution of 200 m x 200 m could not accurately predict NO₂ at traffic locations in
371 urban areas (Liu et al., 2007). It is important to highlight, however, that assessing the
372 very local variability of air pollution in any Spanish city is beyond the scope of the
373 CALIOPE modeling system.

374 Estimating long-term exposures to air pollution in population-based studies is always
375 challenging and requires considerable efforts in terms of time and resources. In order to
376 minimize the exposure misclassification resultant from the use of central monitoring
377 sites to assign community-level exposures, several exposure assessment methodologies
378 are being increasingly applied in cohort studies, such as interpolation methods, LUR
379 models, or dispersion models, each of them providing a different degree of
380 spatiotemporal resolution (Jerrett et al., 2005). LUR modeling is probably the most used
381 approach in the last years because it provides a detailed spatial resolution at a relatively
382 low cost, but its poor temporal resolution poses a problem when attempting to estimate
383 chronic exposures to air pollution over long periods of time. As a result, a few recent
384 studies have combined LUR models with output data from dispersion models, satellite-
385 based data, or meteorological data in order to incorporate long-term temporal variation
386 (Kloog et al., 2011; Liu et al., 2012; Molter et al., 2010). Despite the limited capability

387 of the CALIOPE modeling system for assigning individual exposures in
388 epidemiological studies at the intraurban level, it is able to capture the spatiotemporal
389 pattern in air pollution concentrations at the regional scale and therefore shows great
390 potential as a complementary tool for assigning long-term exposures to air pollution in
391 epidemiologic studies at this scale. The ability to estimate concentrations for any time
392 length makes it particularly appealing for the investigation of the contribution of health
393 effects on different shorter and intermediate time scales. In particular, in pregnancy
394 cohorts one could easily derive exposure terms for any time period.

395

396 **5. Conclusions**

397

398 The present study has assessed the performance of the CALIOPE air quality forecasting
399 system over a large sample of 4-week NO₂ measurements taken with passive samplers
400 in the province of Girona within the framework of an epidemiological study. The
401 evaluation over the study area showed that the model is able to reproduce the spatial
402 variability of 4-week NO₂ concentrations, with a correlation coefficient of 0.78 and a
403 RMSE of 9.8 µg/m³. Results also showed that the performance of the modeling system
404 depends on the type of subarea, and is affected by the sub-grid emission sources,
405 particularly in urban settings. Despite its limitations in simulating within-city contrasts
406 in air pollution concentrations, this study strongly supports the use of CALIOPE output
407 data as a valuable and complementary tool to incorporate regional spatial variability and
408 long-term temporal variability in epidemiologic studies.

409

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Table 1. Descriptive statistics of measured and modeled NO₂ concentrations in the whole REGICOR study area and by groups of towns. Means, standard deviations (SD), and root mean square errors (RMSE) are shown in µg/m³

Area	No. of tubes	No. of measurements^a	Measured mean (SD)	Modeled mean (SD)	Modeled mean after correction^b (SD)	r^c	RMSE^d
All REGICOR towns	884	165	29.3 (15.0)	6.8 (3.5)	27.7 (14.3)	0.78	9.8
Girona-Salt	534	47	46.0 (12.9)	10.6 (2.8)	43.1 (11.3)	0.58	11.3
Banyoles-Porqueres	47	26	20.8 (7.1)	3.7 (0.9)	15.1 (3.8)	0.68	7.7
Blanes	141	20	28.2 (10.7)	7.8 (2.0)	31.7 (8.1)	0.40	10.8
La Bisbal-Palafrugell-Palamós	110	29	25.7 (12.8)	6.4 (3.0)	26.2 (12.2)	0.63	10.5
Olot-Sant Joan	34	26	21.0 (7.2)	3.6 (1.1)	14.9 (4.5)	0.45	8.9
Llagostera-Sta. Cristina	18	17	16.2 (4.8)	5.4 (1.8)	21.9 (7.4)	0.76	6.7

^a After averaging tubes located within the same 4 km x 4 km cell during the same sampling campaign

^b After adjusting the modeled mean by the annual modeled to measured ratio at the fixed monitoring station

^c Pearson correlation coefficient

^d Computed with the corrected modeled concentrations

Figure 1. Map of the REGICOR study area including the location of NO₂ sampling sites and study participants

Figure 2. Annual average concentrations of NO₂ (μg/m³) for the study period simulated by CALIOPE at a 4 km x 4 km spatial resolution

Figure 3. Four-week modeled vs. measured concentrations of NO₂ (μg/m³) at the fixed urban background monitoring station of Girona city over the 1-year study period. $y=16.0 + 0.6x$ ($R^2=0.19$). The scatter plot includes the 1:1 reference line. Modeled concentrations are corrected by the modeled to measured ratio obtained at this station for the whole period.

Figure 4. Modeled vs. measured concentrations of NO₂ (μg/m³) at the REGICOR sampling sites during a 4-weeks period. Measurements conducted during the same sampling campaign at all the sites located within the same grid cell are averaged. In (B), the regression line is weighted by the number of measurements used for computing the mean value at each sampling site, and the size of the dots and triangles is proportional to this number. $y=5.8 + 0.7x$ ($R^2=0.61$) in (A), and $y=17.5 + 0.5x$ ($R^2=0.33$) in (B). The scatter plots include the 1:1 reference line. Modeled concentrations are corrected by the annual modeled to measured ratio at the fixed monitoring station.

Figure 1
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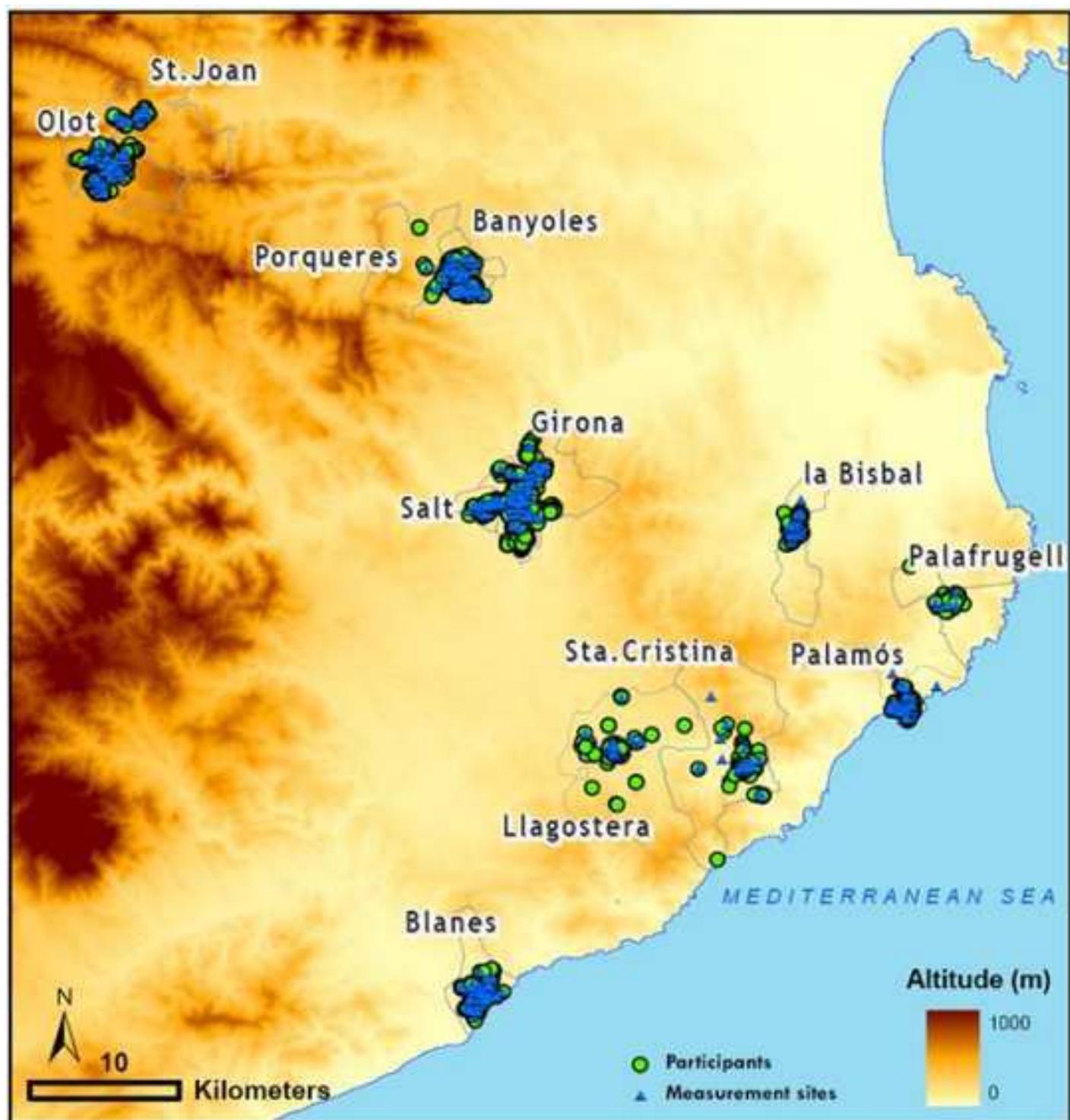


Figure 2

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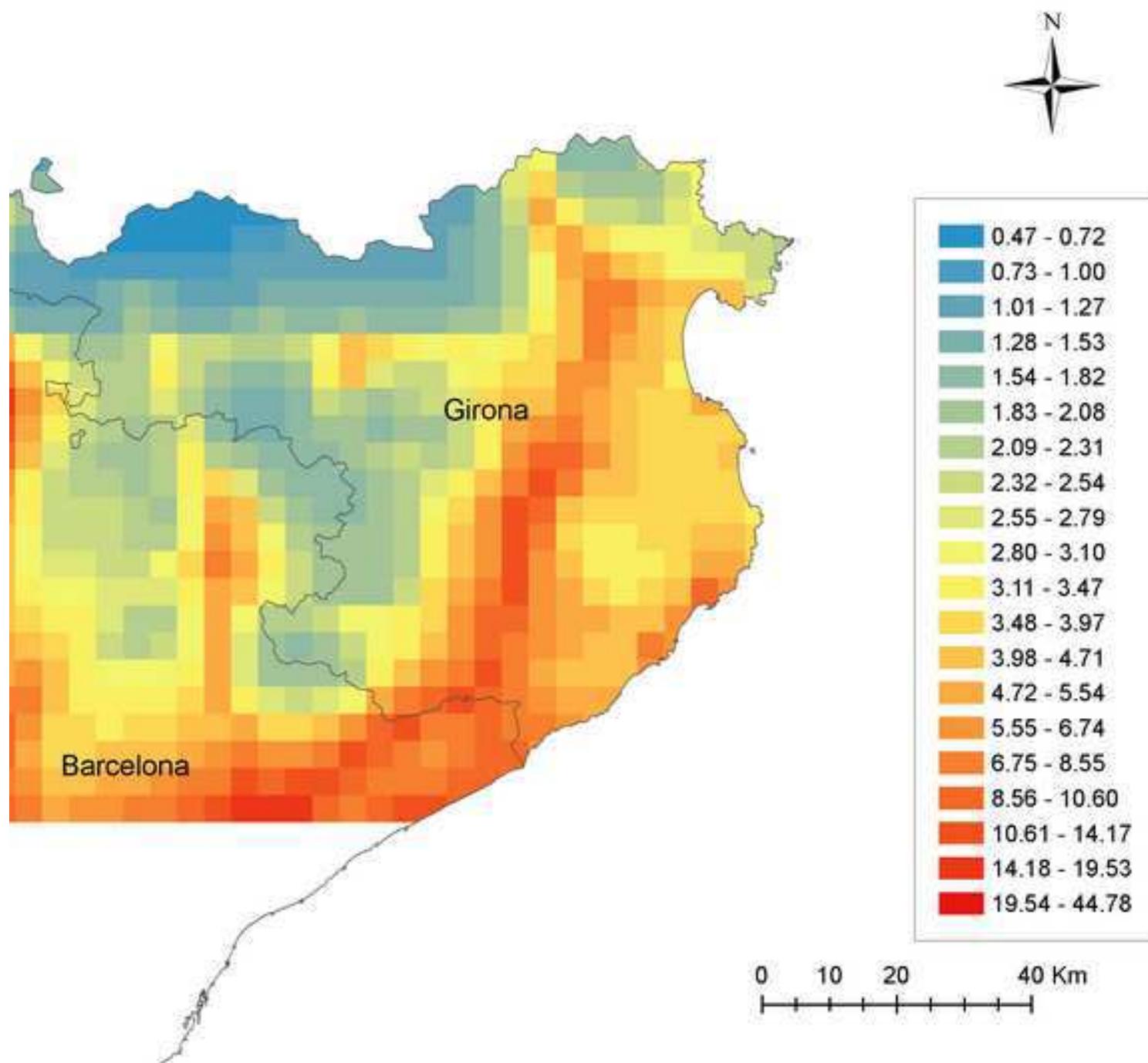


Figure 3

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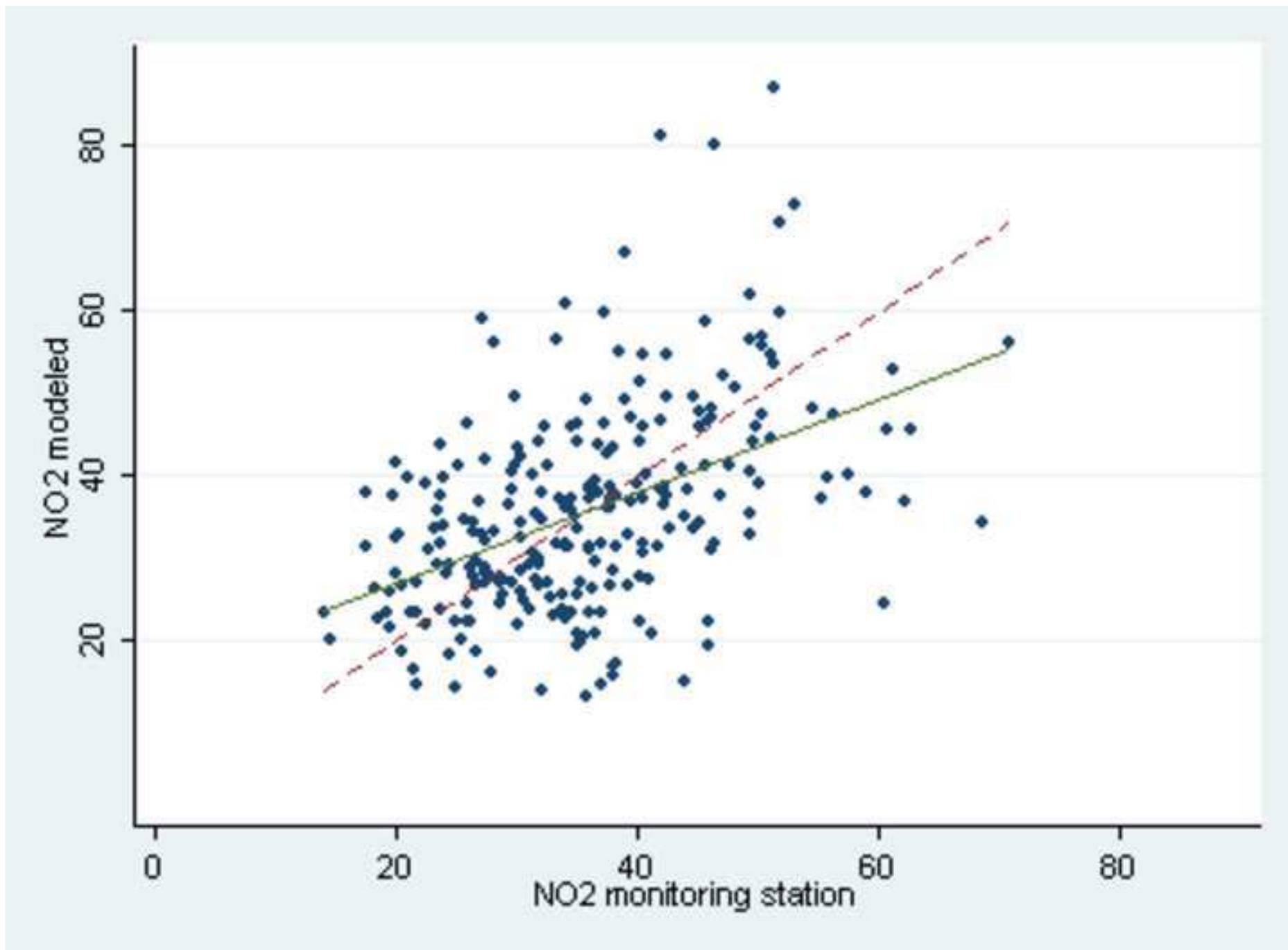


Figure 4

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