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Hernández Crespo, C.; Martín Monerris, M. (2015). Determination of background levels and pollution assessment for seven metals (Cd, Cu, Ni, Pb, Zn, Fe, Mn) in sediments of a Mediterranean coastal lagoon. CATENA. 133:206-214. doi:10.1016/j.catena.2015.05.013.



The final publication is available at

<http://dx.doi.org/10.1016/j.catena.2015.05.013>

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Additional Information

Manuscript Number: CATENA3919R2

Title: Determination of background levels and pollution assessment for seven metals (Cd, Cu, Ni, Pb, Zn, Fe, Mn) in sediments of a Mediterranean coastal lagoon.

Article Type: Research Paper

Keywords: background levels; Albufera de Valencia; sediments; metals; sediment cores

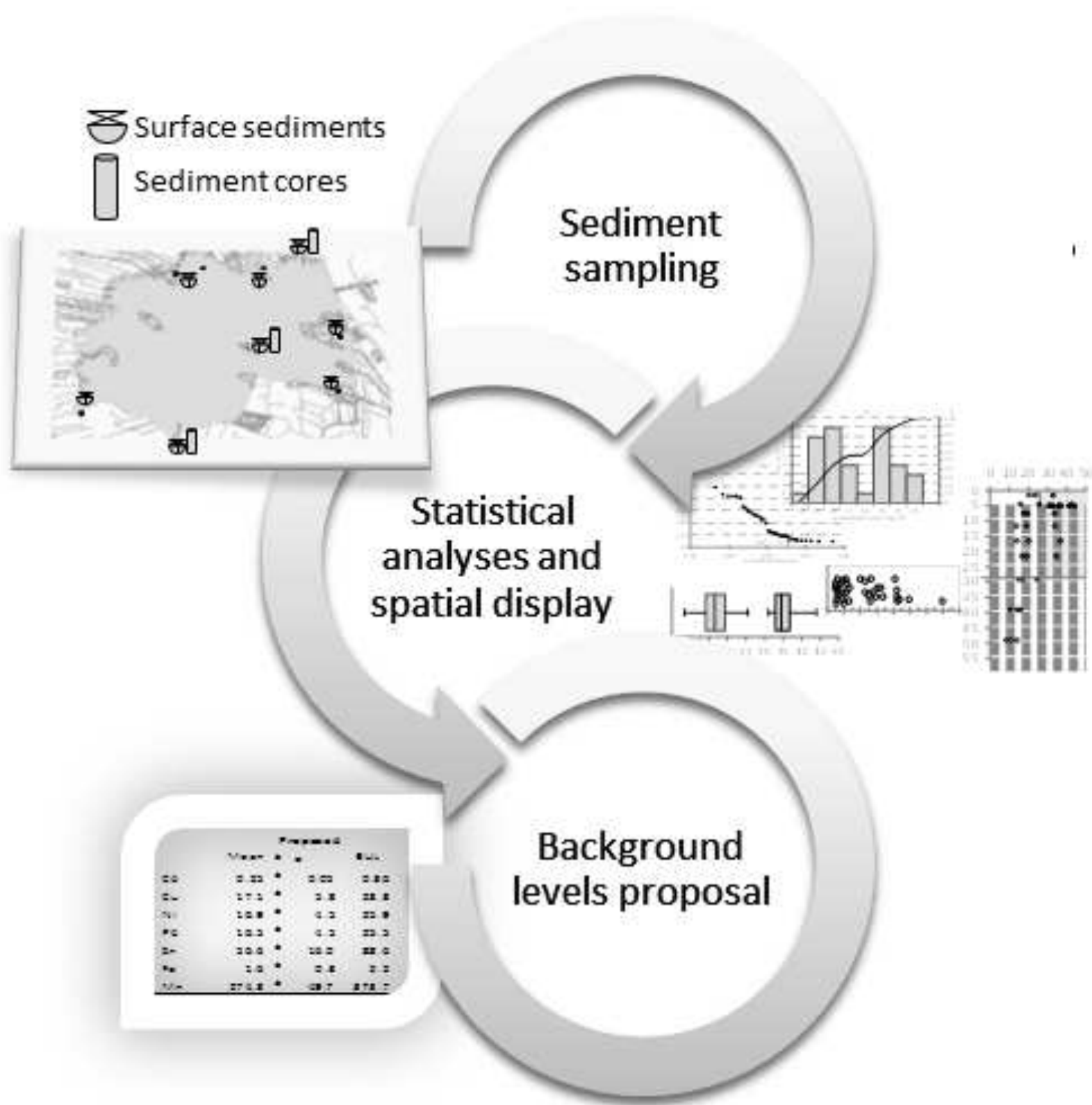
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Abstract: The determination of background levels of trace metals in soils and sediments is a key point for the proper assessment of pollution degree. This study demonstrates the suitability of integrating geological and statistical methods for the reliable determination of background levels, applying it to the sediments of Lake Albufera, a wetland of international importance that is highly eutrophic. The procedure followed includes sampling of sediment cores at different points of the lake, including reference sites, and the subsequent statistical analysis of the data, comprising descriptive statistics, probabilistic plots and modal analysis. The final proposal of background levels considers the data subset separated by the statistical analysis and the spatial and age characteristics of sediments, proving the usefulness of jointly using geological and statistical methods. The upper limits of the background populations, defined as the mean+2 σ and expressed in mg/kg, are Cd (0.38), Cu (28.8), Ni (25.9), Pb (25.5), Zn (88.6), Fe (2.2%) and Mn (345.7). Background levels proposed for different parts of Spain, found in an extensive literature review, are also provided in this article. Once determined the background levels, the assessment of pollution degree of sediments using pollution indexes indicates that the top 25 to 30 cm of sediments has a pollution level between moderate and severe in the peripheral sites, which are nearest to the pathways of contamination, and that the north zone of the lake is the most polluted by the group of the five metals, including Cd, Cu, Ni, Pb and Zn.



Highlights:

- We establish background levels for metals to sediments of Lake Albufera.
- An integrated method of geological and statistical approaches has been used.
- The depth to which sediments are contaminated is provided, the upper 30 cm.

1 Determination of background levels and pollution assessment for seven metals (Cd, Cu, Ni, Pb, Zn, Fe,
2 Mn) in sediments of a Mediterranean coastal lagoon.

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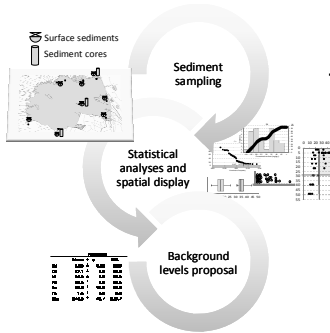
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26 north zone of the lake is the most polluted by the group of the five metals, including Cd, Cu, Ni, Pb
27 and Zn.

28 Graphical abstract



29

30 1. Introduction

31 Trace metals can be found in soils and sediments without human impact, having concentrations
32 derived from rock weathering. These natural concentrations are commonly called “background
33 levels” in environmental studies. Many human activities introduce metals into the environment (e.g.,
34 mining, traffic, agriculture, industries, wastewater treatment plants –WWTP–, waste landfills), thus
35 increasing the concentrations of metals in soils and sediments above the background levels. In lentic
36 water bodies, the tendency of metals to bind to the solid phase promotes their removal from the
37 water column by sedimentation, and thus their accumulation in sediments. Therefore, it is important
38 to know the background concentrations to assess the pollution status of sediments, which is also
39 necessary to evaluate the ecological status of water bodies according to the European water
40 framework directive (WFD). Regarding the background level or concentration concept, it is important
41 not to confuse it with the term “baseline value”. The latter refers to the levels currently measured
42 and serves to quantify future changes (Reimann and Garret, 2005).

43 Different approaches have been developed to determine geochemical background concentrations.
44 Methods are usually classified into direct (empirical or geochemical) or indirect (statistical or
45 theoretical), and both can be combined, leading to integrated methods (Dung et al., 2013; Galuzska
46 and Migaszewski, 2011). Within each category, several procedures can be found:

47 • Direct methods. In these methods, the background concentrations are obtained by analysing
48 samples representing the pre-industrial era or pristine areas.

49 – Historical aspect. The background concentration is estimated as the mean or median of
50 samples representing the pre-industrial period, such as deep sediments or deep soil
51 horizons, glacial ice cores, archival plants from herbaria or tree rings (Galuzska and
52 Migaszewski, 2011).

53 – Contemporary aspect. In this case, it is estimated as the mean or median of samples
54 collected from pristine areas, far away from pollution sources.

55 • Indirect methods or statistical approaches. These methods consist of sampling a large number of
56 sites and using statistical tools and spatial analysis to separate, within a data set, the background
57 concentration from that related to anthropogenic sources. Samples identified as polluted can be
58 single or multiple outliers, or can represent distinct populations. Several methods of statistical
59 sieving can be used:

60 – Tukey boxplots, which identify as outliers any values beyond the whiskers of the boxplot,
61 where the upper whisker = $\max(x|x < \text{upper inner fence})$ and the upper inner fence = $Q3 + 1.5$
62 $(Q3 - Q1)$, $Q1$ is the first quartile (25% of data) and $Q3$ is the third quartile (75% of data)
63 (Reimann et al., 2005; Rodríguez et al., 2006).

64 – Empirical cumulative distribution functions (ECDF) or probabilistic plots, the latter
65 representing the accumulated data on a normal probability scale. These permit the detection
66 of deviations from normality and the presence of multiple populations by slope changes and
67 breaks in the plot (Reimann et al., 2005). To avoid subjectivity in identifying the inflexion

68 point, it is recommended to identify the inflexion point as the value that yields the minimal
69 skewness for the resulting background population (Tobías et al., 1997; Peris, 2006).

70 – 4σ -outlier test. This requires the removal of potential outliers, identified in the ECDF for
71 example, from the dataset and the calculation of the mean and standard deviation for the
72 remaining sub-set of data. Then, the previously defined outliers can be classified objectively
73 if they are further from the mean than 4σ (Matschullat et al., 2000).

74 – Iterative 2σ -technique. The mean and standard deviation are calculated for the original data
75 set. Subsequently, all of the values beyond the $\text{mean} \pm 2\sigma$ are omitted, and the procedure is
76 repeated until all remaining values lie within this range (Matschullat et al., 2000). This
77 technique is considered to be mathematically less robust than the outlier test by other
78 authors (Matschullat et al., 2000). Nevertheless, it is better at reducing the upper limit than
79 the 4σ -outlier test and the calculated distribution (described below), and yields realistic and
80 plausible values (Roca et al., 2012).

81 – Calculated distribution function. The background distribution function is calculated by
82 “mirroring” every single value lower than the median against the original median value by
83 adding the distance from each value to the median, thus obtaining new values larger than
84 the median (Matschullat et al., 2000).

85 – Maximum likelihood mixture estimation (MLME), also called modal analysis, has been
86 demonstrated to be useful in determining background values for biota and sediments (Carral
87 et al., 1995; Rodríguez et al., 2006). This technique decomposes a multi-mode distribution
88 function into several normal distributions. The sub-populations are centred on the modal
89 values supplied by a previous identification technique. Usually the NORMSEP routine is
90 employed, where the sub-populations are actually different if the separation index
91 ($\Delta\text{mean}/\Delta\sigma$) is higher than 2 (Carral et al., 1995).

92 – Linear regression between the concentration of an element and one or several conservative
93 factors (e.g., fine fraction, Fe, Al) considered as inert or not influenced by anthropogenic

94 activities. This tool also allows the identification of values not belonging to the background
95 population, such as those that fall beyond the confidence interval (95%) (Dung et al., 2013).

96 – Principal component analysis (PCA) and cluster analysis are also statistical techniques useful
97 for identifying relationships between metals and other variables, and for grouping different
98 populations from a dataset (Rubio et al., 2000; Aloupi and Angelidis, 2001; Micó et al., 2006;
99 Blasco et al., 2010; Esmaeli et al., 2014).

100 Once the background population has been separated, the values representing this population
101 should be expressed as a range, which has traditionally been defined as the mean \pm two times
102 the standard deviation ($\text{mean} \pm 2 \cdot \sigma$). However, Reimann et al. (2005) demonstrated that the
103 median \pm two times the median of the absolute deviations ($\text{median} \pm 2 \cdot \text{MAD}$) is a more
104 appropriate estimator, even for normal populations.

105 • Integrated methods combine several direct and indirect approaches. Several authors have
106 demonstrated the convenience of integrating different methods to provide more reliable
107 background thresholds. For instance, Galuzska (2007) sampled clean areas and used statistical
108 analysis, Reimann et al. (2005) proposed a heuristic consisting of sampling different sites and
109 applying statistical and spatial analyses, and Matschullat et al. (2000) applied different statistical
110 tools to data obtained from drill sediment cores.

111 The use of direct methods is criticized for various reasons: sampling of deep sediment is considered
112 technically difficult and expensive (Carral et al., 1995); it requires expert knowledge of the sampling
113 area and about the geochemical behaviour of the investigated elements, and there may be
114 subjectivity in selecting samples (Matschullat et al., 2000; Galuzska and Migaszewski, 2011).
115 Sampling of deep sediment is also questionable due to the eventual depletion of some elements
116 because of their natural properties, rather than owing to a lack of anthropogenic pollution (Reimann
117 and Caritat, 2005) and it pre-supposes that there have been no post-depositional movements (Carral
118 et al., 1995). Nevertheless, it is a method that is amply used (Blasco et al., 2000; Cobelo-García et al.,

119 2003; Tylmann et al., 2011), and in some cases it is relatively easy to take long sediment cores (i.e.,
120 shallow waterbodies). Despite the above drawbacks, this method is advantageous in the sense that
121 the results represent actual data and are not subject to any processing. On the other hand, the
122 sampling of pristine areas is accompanied by the uncertainty of whether they are indeed free of
123 anthropogenic pollution (Galuzska and Migaszewski, 2011).

124 This study aims to (1) establish background levels for seven metals (Cd, Cu, Ni, Pb, Zn, Fe, Mn) by
125 integrating different methods for sediments from Lake Albufera de Valencia (Spain), a wetland of
126 international importance according to the Ramsar Convention, (2) assess the extent of surface
127 sediment pollution, and (3) compare the proposed background levels with those proposed for other
128 locations in Spain and perform a review of studies of background concentrations for soils and
129 sediments.

130 **2. Materials and Methods**

131 2.1 Study area

132 Lake Albufera is a coastal lagoon located 10 km southeast of Valencia (Spain). It has an area of 2400
133 ha, a mean water depth of 1 m and an average sediment thickness of 70 cm. The climate is
134 Mediterranean, with a low mean annual precipitation (551 mm) and intense storms during autumn
135 (up to 100-300 mm/d). The water temperature of the lake varies between 11°C (Dec-Jan) and 28°C
136 (Jul-Aug). Since the 1970s, the lake has been highly eutrophic (with an annual mean chlorophyll- α
137 concentration over 100 $\mu\text{g/L}$) due to several anthropic pressures (urban, industrial and agricultural).
138 Sediments are mainly silty clay, with high contents of organic matter and metals in the layers close to
139 the surface. So far, the concentration of acid volatile sulphide has been sufficient to retain metals as
140 metal sulphides, but the concentration has a decreasing trend (Hernández-Crespo and Martín, 2013).
141 Thus, it is important to establish the background levels to assess the degree of sediment metal
142 contamination.

143 2.2 Sampling and analytical determinations

144 Procedures for sampling and chemical analysis are described in detail in Hernández-Crespo et al.
145 (2012, 2013). Briefly, 9 sites inside the lake (identified as 1-7, 10-11 in Fig. 1) were selected for
146 surface sediment sampling (Sep 2008); among these, sites 1, 6 and 11 were selected for sediment
147 core sampling (Sep 2011 and Mar 2012). The peripheral sites 1 and 11 represent areas with higher
148 contamination, while site 6, located in the central area of the lake, is considered as a reference site
149 that is less affected by pollution inputs. With the aim of obtaining vertical profiles, the sediment
150 cores were sectioned into 9 slices of increasing thickness (3x3 cm, 3x5 cm, 3x10 cm). The three
151 sampling campaigns provided a total of 63 samples. Nevertheless, the average values of the two
152 samplings of sediment cores have been utilized for statistical analyses because the results did not
153 differ significantly ($p>0.05$), thus leaving 27 samples. Gathering the surface (9) and sediment cores
154 (27), a total of 36 samples are available for statistical analysis. Metal content was determined by
155 extraction with aqua regia and quantification by atomic absorption spectrometry (AAS). Quality
156 control was performed by analysing blanks and certified reference material (CRM-320), with the
157 following concentrations and recoveries: Cd (0.47 mg/kg; 98%), Cu (42.0 mg/kg; 100%), Ni (55.0
158 mg/kg; 95%), Pb (30.0 mg/kg; 93%), Zn (124.4 mg/kg; 110%), and Fe (3.3%; 99%).

159 2.3 Background assessment

160 An integrated method, including direct (geochemical) and indirect (statistical) approaches, has been
161 applied to estimate the background concentrations for seven metals (Cd, Cu, Ni, Pb, Zn, Fe, Mn),
162 approximately following the heuristic proposed by Reimann et al. (2005) for data inspection and
163 selection of background thresholds. The steps taken are: (1) Core sediment sampling, to reach
164 unpolluted layers deposited before urban and industrial development of the surrounding area
165 (Hernández-Crespo and Martín, 2013). (2) Probabilistic plots and skewness determination to perform
166 a first inspection and separation of the dataset. (3) Boxplots, histograms and scatterplots are
167 displayed to make a deeper inspection. (4) A MLME is realized, employing FISATII software to apply

168 the modal progression analysis (Gayanilo et al., 2005). This analysis consists of two stages: first, a
169 decomposition of the distribution into their components to identify potential means, using the
170 Bhattacharya's routine; and second, a refinement of results with the NORMSEP routine. Once the
171 populations are separated, new boxplots showing the different populations are represented. (5)
172 Spatial plots, in the form of vertical profiles, are used to support the separation performed with the
173 above techniques. (6) Finally, statistical estimators (mean \pm 2 σ and median \pm 2 MAD) are calculated to
174 define the background population.

175 2.4 Pollution assessment

176 The pollution degree has been assessed, with the calculation of three widely used indexes (Dung et
177 al., 2013) and taking into account the upper limit of background concentration determined as
178 described above. The indexes calculated are:

179 Enrichment factor:
$$EF = \frac{\frac{C_{metal_sample}}{C_{normalizer_sample}}}{\frac{C_{metal_background}}{C_{normalizer_background}}}$$

180 Geoaccumulation index:
$$I_{geo} = \text{Log}_2 \left(\frac{C_{metal_sample}}{1.5 \cdot C_{metal_background}} \right)$$

181 Pollution Load index:
$$PLI = \left(\frac{C_{metal_1_sample}}{C_{metal_1_background}} \cdot \dots \cdot \frac{C_{metal_n_sample}}{C_{metal_n_background}} \right)^{\frac{1}{n}}$$

182 The enrichment factor divides the concentration of metal, in the sample and the background, by a
183 normalizer element. Several constituents can be used as normalizers (Al, Fe, organic matter, clay
184 fraction, etc.) (Rubio et al., 2000). In this study, the normalizer used was Fe because it did not show
185 evidence of an anthropogenic source, as discussed in section 3.1.

186 3. Results and Discussion

187 3.1 Background determination

188 According to the stages described in section 2.3, a first visual inspection was made with probabilistic
189 plots (Fig. 2). The possible break or inflexion points are indicated with arrows on the plot, and the
190 skewness of the population separated by each point is also displayed. As shown in Figure 2, for
191 almost all metals studied, at least one inflexion point was clearly identified, and the data subset
192 selected as the background population was the one with minor skewness (Table 1). For all metals
193 except for iron, the selected sub-population had lower skewness than the population as a whole,
194 suggesting that discrete populations can actually be separated and that iron can be used as a suitable
195 normalizer element because it is not influenced by anthropogenic activities.

196 Subsequently, the complete dataset was subject to further examination by means of histograms,
197 scatter plots and box-and-whisker plots (Fig. 3). Histograms and scatter plots allowed us to observe
198 polymodality for Cd, Cu, Ni, Pb and Zn, and boxplots indicated a wide variability in the data, with the
199 existence of outliers for Cu, Pb and Mn. The existence of multiple populations was demonstrated by
200 the MLME, employing the NORMSEP routine in the FISATII software, which indicates the existence of
201 different sub-populations if the separation index is higher than 2. The sub-populations extracted are
202 shown in Figure 3 by the new histograms and boxplots. The separation suggested by the NORMSEP
203 routine is in agreement with the polymodality previously observed in the histograms and scatter
204 plots, and agrees as well with the division made by the inflexion points in the probabilistic plots for
205 most metals (see Table 1). The $\text{mean} \pm 2\sigma$ range was selected for characterizing the background
206 population obtained by both tools (probabilistic plots and MLME) because the data subset was
207 normally distributed according to the Shapiro-Wilk test. The statistical estimator $\text{median} \pm 2\text{MAD}$ was
208 not selected because, although for most metals (Cd, Cu, Ni, Pb, Mn) the difference between both
209 estimators ($\text{mean} \pm 2\sigma$ and $\text{median} \pm 2\text{MAD}$) was low, the $\text{median} + 2\text{MAD}$ excluded several samples
210 belonging to the populations defined by the statistical approaches and corresponding to aged or
211 reference sediments for Zn and Fe.

212 Furthermore, an estimate of background values with the iterative 2σ -outlier test was performed,
213 giving significantly higher values than the other methods tested. Therefore, the authors considered
214 that this technique would be appropriate for characterizing the background population when applied
215 to samples from uncontaminated areas, but not to identify sub-populations (contaminated and
216 uncontaminated).

217 The upper limits of background values obtained above were represented in vertical profiles to
218 compare them with those obtained by separating the samples according to their age (Fig. 4). For the
219 latter determination, this technique considered the dating of lake sediments performed by Rodrigo
220 et al. (2010). The discrepancy between statistical methods (probabilistic plots and modal analysis)
221 was solved with the help of this information. For the case of Cd, the resultant upper limit from both
222 methods was very similar, so differences were minor. However, in the case of Cu the difference
223 between both methods was important, and the age of sediment helped us to select the more
224 appropriate background upper limit, this being the one provided by the modal analysis, because it
225 included the older sediments and those from the reference site. In Figure 4, the shaded quadrant is
226 delimited vertically by the selected upper limit of the background concentration, determined with
227 probabilistic plots and/or modal analyses, and horizontally by the sediment depth corresponding to
228 an age of approximately 50 years. Thus, the values falling in this shaded quadrant are above
229 background upper limits, and correspond to concentrations measured in the upper layers of the
230 peripheral sampling sites. Values to the left of the upper limit in the surface layers correspond to the
231 central point taken as a reference point, which is less affected by pollution. Additionally, values
232 below 30 cm depth correspond to sediments accumulated before the development of the
233 surrounding towns, which grew enormously since the 1960s. Therefore, it can be concluded that the
234 statistical analysis performed has provided an appropriate segregation of data and highlights the
235 usefulness of vertical profiles for validating this segregation. The integrated use of statistical
236 techniques and the sampling of uncontaminated sites and sediment cores are advisable to define the
237 background levels with a high level of confidence.

238 With the aim of comparing the results obtained for Lake Albufera with background levels proposed
239 elsewhere in Spain, a review of scientific literature was performed. Data obtained from different
240 publications are summarized in Figure 5 and Table 2, where technical information such as the
241 analytical extraction procedure and the statistical approach used are indicated. In Figure 5, it can be
242 observed that the background values proposed in this study are generally in the lower-middle range
243 of those proposed for other sites in Spain. A spatial trend that marks zones with higher or lower
244 metal concentrations than the rest is not observed. The differences among locations may be due to
245 actual differences among soils and sediments but also due to the different extraction procedures and
246 statistical techniques employed. This review indicates that there is significant variation among the
247 methods used, not only in analytical determination or statistical techniques but also in the
248 background definition—usually the background level is defined as the mean, and the upper limit is
249 reserved to propose baseline or reference values—and this makes the comparison of data difficult.
250 Hence, a uniform set of procedures should be agreed on to identify differences related to soil and/or
251 sediment. Another noteworthy aspect is that the majority of published studies analyse soils, and a
252 smaller proportion were aimed at studying the sediments. Therefore, it is considered interesting to
253 observe the differences among studies aimed at jointly analysing soils and sediments of water bodies
254 located in the same watershed.

255 3.1 Pollution assessment

256 Once the background upper limit (or threshold) is established, pollution indexes can be calculated. In
257 this study, the EF, I_{geo} and PLI were calculated. Figure 6 shows the values obtained for the EF and PLI
258 indexes for the metals Cd, Cu, Ni, Pb and Zn. According to the ranking cited in diverse studies (EF \leq 1:
259 no enrichment; 1<EF \leq 3: minor enrichment; 3<EF \leq 5: moderately enrichment; 5<EF \leq 10 moderately
260 severe enrichment; Dung et al., 2013), the level of pollution in the upper 25-30 cm of peripheral sites
261 (sites 1 and 11) was minor (Ni, Pb) or moderate (Cd, Cu, Zn), reaching the moderately severe level for
262 copper and zinc at the second layer (5 cm) of sediments from sites 1 and 11, respectively. I_{geo} was

263 lower in all cases (not shown), classifying the sediment as less polluted than the EF. Other authors
264 have also observed this trend in I_{geo} , where the degree of pollution is minimized because of the
265 numerical factor (1.5) introduced to take into account the possible variations of background levels
266 (Sutherland, 2000; Dung et al., 2013). In this case, as the background level has been defined as an
267 upper limit, the authors consider the use of EF to be more appropriate because the upper limit
268 already takes into account the variability of the background population. Finally, PLI combines all
269 metals into one index, providing an integrated assessment of the pollution degree. According to the
270 PLI results, sediments in the perimeter sites are polluted to the 25-30 cm depth, reaching a higher
271 degree of pollution at site 1, which is consistent with the higher urban pressures found in the north
272 ditches of the lake (Pascual-Aguilar et al., 2013). Average values of organic carbon were 2.7, 3.7 and
273 2.9% OC, and of fine fraction (<63 μm) were 86.0, 97.0 and 93.1%, for sites 1, 6 and 11, respectively
274 (Hernández-Crespo and Martín, 2013). Significant correlations between the concentrations of metals
275 and organic carbon or fine fraction were only found for data from site 11, where both variables were
276 positively correlated with metals, indicating an affinity of metals to organic matter and the fine
277 fraction.

278 According to the WFD classification of ecological status, the condition required for non-synthetic
279 pollutants to reach a 'high status' is that the concentrations remain within the range normally
280 associated with undisturbed conditions or background levels. The category of 'good status' is
281 obtained if concentrations are below the Environmental Quality Standards (EQS) proposed as
282 specified in WFD. The categories of 'moderate status' or worse may be applied if these EQS are
283 exceeded (Rodríguez et al., 2006). EQS for sediments have not yet been adopted in Spain, but
284 sediment quality guidelines (SQG) proposed by other authors (MacDonald et al., 2000) can be used
285 as a reference, whereby the surface sediments are around the threshold effect concentration (TEC)
286 for Cd (0.99 mg/kg), Cu (31.6 mg/kg) and Pb (35.8 mg/kg), and near the probable effect
287 concentration (PEC) for Ni (48.6 mg/kg) at site 1 and Zn (459 mg/kg) at sites 1 and 11. In the
288 subsurface layer (second layer from surface) at site 1, Cu, Ni and Pb are very close to the PEL, and Zn

289 exceeds it. Thus, actions on the sediment should be suggested at the peripheral sites. Nevertheless,
290 these SQG can be used as screening tools to identify hot spots, but it is recommendable to perform
291 biological assays and to take into account sediment properties, such as the organic carbon (OC)
292 content or fine fraction (Campana et al., 2013). If Cu concentrations are normalized with respect to
293 OC (2.7-3.7%, Hernández-Crespo and Martín, 2013) and fine fraction (88-97%, Hernández-Crespo and
294 Martín, 2013), only the subsurface sediment at site 1 (5.6 mg <63 μ m Cu/g OC) exceeds the no-effect
295 value determined by Campana et al. (5.5 mg <63 μ m Cu/g OC). Consequently, further studies are
296 needed to make proper decisions about actions to be performed on the sediments of Lake Albufera.
297 Note that the decrease of the degree of pollution observed in the uppermost layer of these two sites
298 with respect to the layer immediately below (Fig. 6) is almost certainly related to a decreased input
299 of metals in recent years, thanks to the measures implemented to prevent the arrival of pollutants,
300 such as the proper collection and treatment of wastewaters.

301 In addition to those measures aimed at avoiding external loads, several measures focused on
302 nutrient internal loads are being carried out to recover the ecological status of the lake, such as
303 treating the lake water in off-shore constructed wetlands (Martín et al., 2013; Rodrigo et al., 2013),
304 or are being considered by stakeholders, such as sediment dredging. This study provides useful
305 information for designing such measures, for instance the thickness of the contaminated layer of
306 sediment. We suggest a prioritization of areas for action, focusing first on the tributary ditches, which
307 are highly contaminated (Hernández-Crespo et al., 2012), and second on a perimeter belt inside the
308 lake, where other measures softer than dredging, such as phytoremediation, could be applied.

309 **4. Conclusions**

310 The determination of background levels of metals for the sediments of Lake Albufera performed in
311 this study is a significant step toward a proper assessment of the degree of sediments contamination
312 and toward proposing appropriate measures for the recovery of its ecological status. The integrated
313 use of geochemical and statistical methods have been demonstrated to be useful for reliably

314 determining background levels, and this integration of methods has allowed the validation of each
315 other as well. Thus, the authors recommend the use of integrated methods to establish background
316 levels, such as the procedure developed in the present study, especially in shallow lagoons where the
317 extraction of sediment cores does not involve a major effort. The stages to be integrated are: (1)
318 sampling of sediment cores in different sites located near and far from the mouth of pathways for
319 pollutants, (2) to analyse and date the sediments, (3) process the results using several statistical
320 methods of sieving to identify different data subsets (contaminated and non-contaminated) and (4)
321 to define the background values considering the information from both previous stages and using the
322 best suited statistical indicators.

323 The determination of background levels lets us know, from now on, the extent of sediment
324 contamination, currently being moderate or less (for Cd, Ni, Pb, Fe, Mn) up to 30 cm deep, and
325 moderately severe for copper and zinc in the surface layers (5 cm) of peripheral sites. The study of
326 the vertical profile of sediments provides information about the thickness of the polluted sediment,
327 which is valuable for the design of measures focused on sediment, such as dredging or other
328 remediation measures.

329

330 **Acknowledgments**

331 The authors gratefully acknowledge the partial funding by the support programme for research and
332 development, first research projects (PAID-06-08-3155) at the Universitat Politècnica de València.

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444 Table 1. Descriptive statistics of background populations according to the separation performed with
 445 probabilistic plots and modal analysis, and final proposed values (mg/kg; except Fe, %). N: subpopulation size. *p*
 446 (S-W): significance level of Saphiro-Wilk normality test. BUL: background upper limit (Mean+2σ).

	Probabilistic plots					Modal analysis					Proposed
	N	<i>p</i> (S-W)	Mean	± σ	BUL	N	<i>p</i> (S-W)	Mean	± σ	BUL	BUL
Cd	13	0.944	0.26	± 0.06	0.38	12	0.746	0.25	± 0.05	0.36	0.38
Cu	8	0.894	11.5	± 1.3	14.2	18	0.097	17.1	± 5.8	28.8	28.8
Ni	21	0.847	16.9	± 4.5	25.9	21	0.847	16.9	± 4.5	25.9	25.9
Pb	20	0.428	16.5	± 4.5	25.5	20	0.428	16.5	± 4.5	25.5	25.5
Zn	15	0.166	56.6	± 16.0	88.6	15	0.166	56.6	± 16.0	88.6	88.6
Fe	36	0.374	1.6	± 0.3	2.2	36	0.374	1.6	± 0.3	2.2	2.2
Mn	34	0.654	266.7	± 39.5	345.7	34	0.654	266.7	± 39.5	345.7	345.7

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Table 2. Review of background values proposed for different sites in Spain. GM: geometric mean, GSD: geometric standard deviation.

Authors	Place	Analytical method	Background estimation	Concentration (mg/kg; except Fe in %)								
				Cd	Cr	Cu	Hg	Ni	Pb	Zn	Fe	Mn
Carral et al., 1995	Galicia (coast)	HNO ₃ +HF	Sediment surface samples - Modal analysis	Upper limit (CI95)	32	28		32	53	122	2.6	275
Cobelo-García et al., 2003	Galicia (Ría Ferrol)	HNO ₃ +HF	Sediment core sampling - Linear regression	Mean±σ	63±14	12±3		26±10	27±7	55±11	2.4±0.5	
Rubio et al., 2000	Galicia (Rías Bajas, Vigo)	HNO ₃ +HF+HClO ₄	Sediment surface samples - PCA	Mean	34.0	29.4		30.3	51.3	105.3	3.5	244.3
Rodríguez et al., 2006	Basque country (coast)	HNO ₃ +HCl	Sediment surface samples - Modal analysis, geographical display	Upper limit (mean+2.7σ)	0.45	71	64	57	66	248	5.4	447
Tobías et al., 1997	Cataluña	X-ray fluorescence	Soil profiles - Probability plots	Antilog(GM+2·GSD)		275	145	83	91	326		
Navas et al., 2002	Aragón	HNO ₃ +H ₂ O ₂	Soil deep samples - Descriptive statistics, geographical display	Median	0.46	19.2	8.4	17.4	5.9	48.2	1.6	284
De Miguel et al., 2002	Madrid	HNO ₃ +HCl	Soil surface samples - descriptive statistics	Upper limit (p95)	0.30	34	20	0.09	20	31	75	710
Sánchez, 2003	Valladolid (M. Campo)	HNO ₃	Soil profile - Descriptive statistics, PCA, cluster	Upper limit (mean+2σ)	0.44	16.1	9.4	9.8	13.8	33.4		
Jiménez et al., 2010	Castilla la Mancha	X-ray fluorescence	Soil profile - Descriptive statistics	Upper limit (mean+2σ)		113.4	27	42.6	44.2	86.5		
Peris, 2006	Castellón	HNO ₃ +HCl	Soil surface samples - Probability plots, descriptive statistics	Upper limit (mean+2σ)	0.63	42.9	50.4	29.2	51.3	154.9	2.6	517
This study	Valencia	HNO ₃ +HCl	Sediment core sampling - Probability plots, modal analysis, descriptive statistics	Upper limit (mean+2σ)	0.38		28.8	25.9	25.5	88.6	2.2	345.7
Micó et al., 2007	Alicante	HNO ₃ +HCl	Soil surface samples - Descriptive statistics	Upper limit (mean+2σ)	0.7	36	28	31	28	83	2.0	402
Roca-Pérez et al., 2010	Eastern Spain	HNO ₃ +HF+HCl+H ₂ O ₂	Soil surface samples natural soils - descriptive statistics	Upper limit (GM/GSD ² , GM·GSD ²)	0.97	217	46	50	137	246	12.4	2691
Pérez-Sirvent et al., 2009	Murcia (Campo de Cartagena)	HNO ₃ +HF	Soil surface samples - PCA, cluster and discriminant analysis	Median (p95)	0.30 (0.43)			0.4 (1.5)		40 (93.5)		
Hernández-Bastida et al., 2005	Murcia (Campo de Cartagena)	HNO ₃ +HF	Soil surface samples - Descriptive statistics, geographical display	Upper limit (median+2σ)	0.53	89	40	38	44	105		628

Table 2. Review of background values proposed for different sites in Spain (cont.).

Authors	Place	Analytical method	Background estimation	Concentration (mg/kg)											
				Cd	Cr	Cu	Hg	Ni	Pb	Zn	Fe	Mn			
Sierra et al., 2007	Almería	HNO ₃ +HF+HClO ₄	Soil profile - ECFD	0.3	123	47.2	1	70.9	93.9	129					
Díez et al., 2009	Granada	HNO ₃ +HCl+HF	Soil profile - ECFD		110	40		56	50	135					
Blasco et al., 2000	Huelva (Odiel estuary)	HNO ₃ +HCl+HF	Sediment core sampling - descriptive statistics		112	1975								11.4	346.1
	Cádiz (Bay Cádiz) Cádiz (Barbate estuary)				97.4	21.3								4.2	402.8
Galán et al., 2008	Andalucía (Ossa-Morena Zone)	HNO ₃ +HF+HCl+HClO ₄	Soil profile - Descriptive statistics		60.4	14.8								2.3	278.0
		HClO ₄	Upper limit (p95)	182	143			73	200	327					

Figure 1. Location of Lake Albufera and sampling sites.

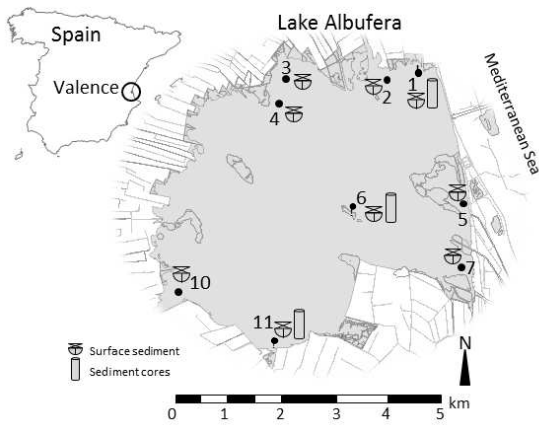


Figure 2. Probabilistic plots for the original dataset. Arrows indicate possible inflexion points, and the values are the skewness coefficients associated with the subpopulations separated by each inflexion point.

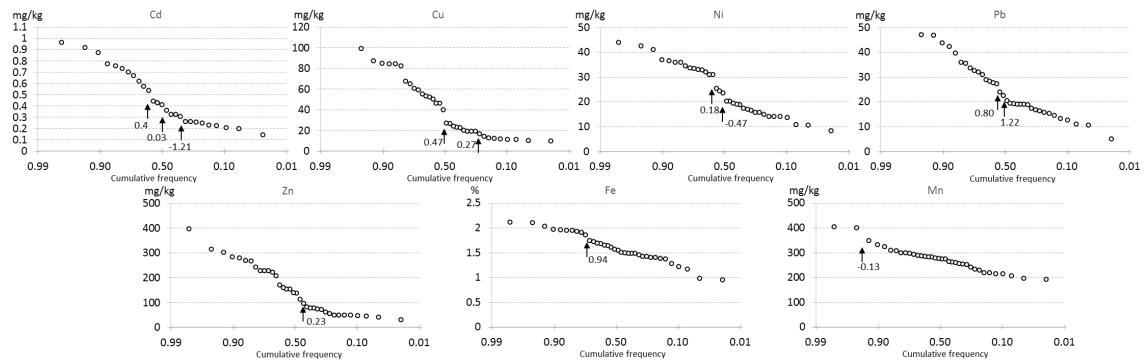


Figure 3. Top down: (a) histograms, scatter plots and box-whisker plots for the original dataset, (b) histograms and box-whisker plots for normal distribution of data subsets extracted by modal analysis.

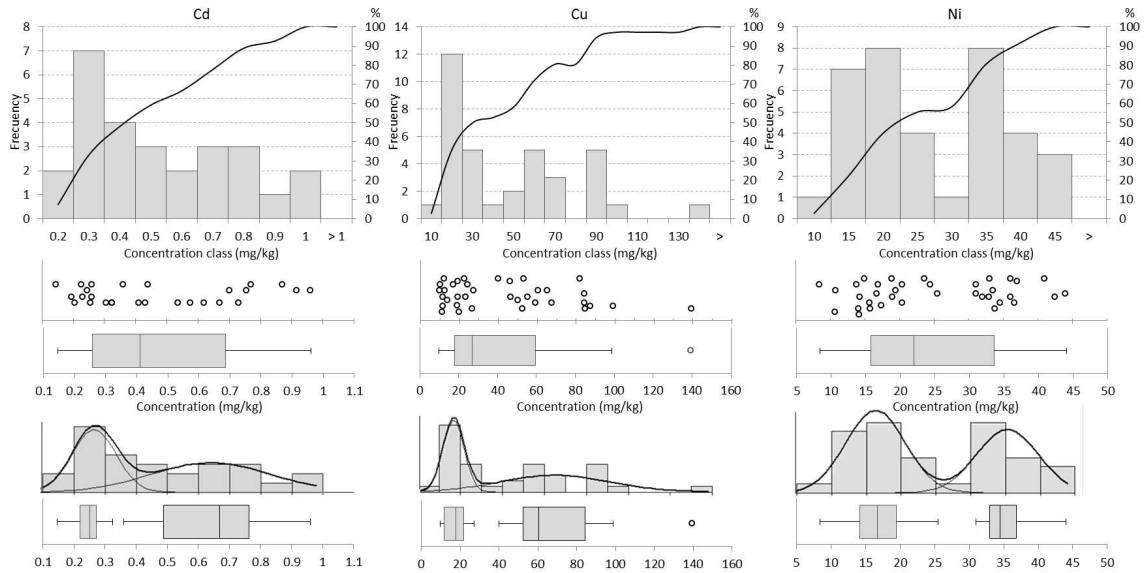


Figure 3 (cont.). Top down: (a) histograms, scatter plots and box-whisker plots for the original dataset, (b) histograms and box-whisker plots for normal distribution of data subsets extracted by modal analysis.

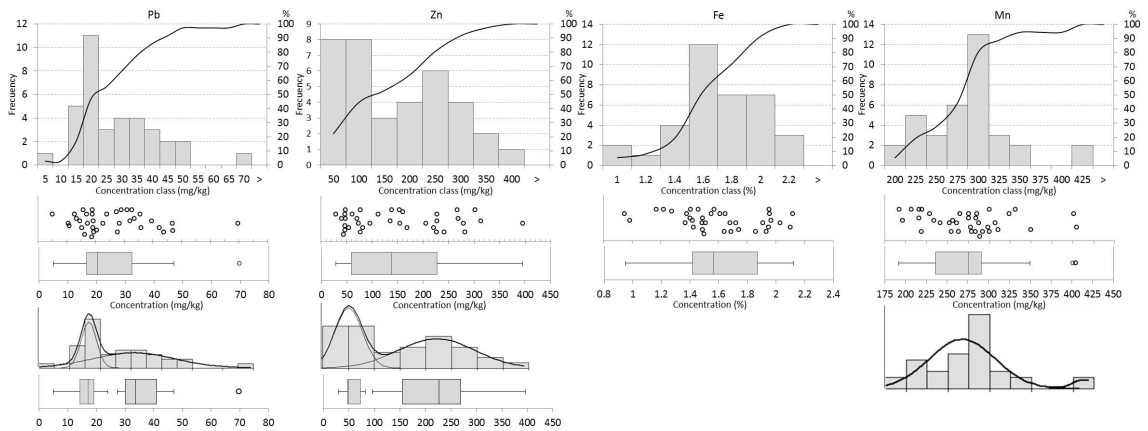


Figure 4. Vertical profiles of sediment age and concentrations of metals. Data for sediment age is adapted from Rodrigo et al. (2010). Vertical lines represent the upper limit of background levels (BUL) obtained by statistical approaches. Horizontal solid lines represent the depth below which the sediment was deposited prior to urban and industrial development.

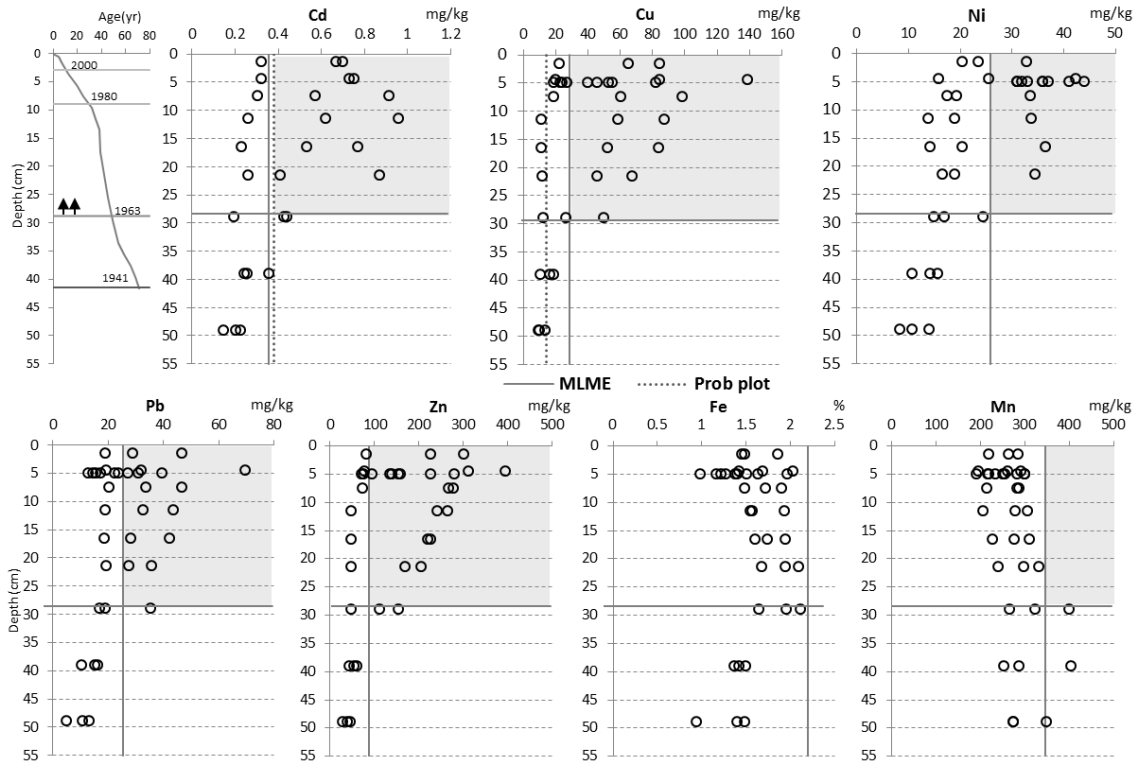


Figure 5. Spatial representation of background values proposed in different locations in Spain. The legend indicates the upper limit of concentration, represented by each size of circle (mg/kg, except Fe in %).

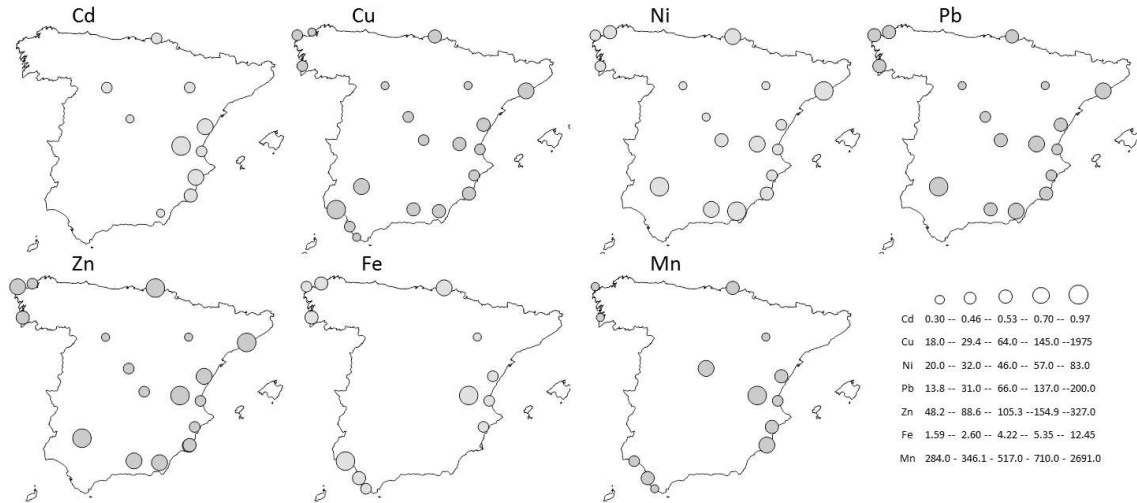


Figure 6. Vertical profiles of Enrichment Factor of sites sampled in Lake Albufera. Results of surface samples are represented by open diamonds. EF scale: $EF \leq 1$: no enrichment; $1 < EF \leq 3$: minor enrichment; $3 < EF \leq 5$: moderately enrichment; $5 < EF \leq 10$ moderately severe enrichment.

