

Comissão Científica

Deolinda Flores  
Helena Sant'Ovaia  
José Brilha  
Manuel Collares Pereira  
Maria João Costa  
Mourad Bezzeghoud  
Rui Salgado  
Teresa Valente



# Jornadas do ICT 2019

24 e 25 de maio de 2019, Auditório do Colégio do Espírito Santo, Universidade de Évora

## Livro de Resumos



Comissão Organizadora

Cláudia Cruz  
Gonçalo Rodrigues  
Joana Ribeiro  
Miguel Maia  
Miguel Potes  
Noel Moreira  
Patrícia Gomes  
Rui Oliveira  
Sara Pereira



Instituto de Ciências da Terra



UNIVERSIDADE DE ÉVORA



Universidade do Minho



## Developing a new Bayesian Risk Index for risk evaluation of soil contamination

Albuquerque MTD<sup>1</sup>, Gerassis S<sup>2</sup>, Sierra C<sup>3</sup>, Taboada J<sup>2</sup>, Martín, JE<sup>2</sup>, Antunes IMHR<sup>4</sup>, Gallego, JR<sup>3</sup>

<sup>1</sup> Instituto Politécnico de Castelo Branco, Instituto de Ciências da Terra, Pólo1, Universidade de Évora, teresa@ipcb.pt

<sup>2</sup> Department of Natural Resources and Environmental Engineering, Univ. of Vigo, Vigo, Spain

<sup>3</sup>INDUROT and Environmental Technology, Biotechnology, and Geochemistry Group, Universidad de Oviedo, Campus de Mieres, Asturias, Spain

<sup>4</sup>Universidade do Minho, Instituto de Ciências da Terra, Pólo2, Universidade do Minho

### Abstract

Soil quality is heavily constrained by industrial and agricultural activities. Potentially Toxic Elements (PTEs) are a threat to public health and the environment alike. In this regard, the identification of areas that require remediation is crucial. In the herein research a geochemical dataset (230 samples) comprising 14 elements (Cu, Pb, Zn, Ag, Ni, Mn, Fe, As, Cd, V, Cr, Ti, Al and S) was gathered throughout eight different zones distinguished by their main activity, namely, recreational, agriculture/livestock and heavy industry in the Avilés Estuary (North of Spain). Then a stratified systematic sampling method was used at short, medium, and long distances from each zone to obtain a representative picture of the total variability of the selected attributes. The information was then combined in four risk classes (Low, Moderate, High, Remediation) following reference values from several sediment quality guidelines (SQGs). A Bayesian analysis, inferred for each zone, allowed the characterization of PTEs correlations, the unsupervised learning network technique proving to be the best fit. Based on the Bayesian network structure obtained, Pb, As and Mn were selected as key contamination parameters. For these 3 elements, the conditional probability obtained was allocated to each observed point, and a simple, direct index (Bayesian Risk Index-BRI) was constructed as a linear rating of the pre-defined risk classes weighted by the previously obtained probability. Finally, the BRI underwent geostatistical modeling. One hundred Sequential Gaussian Simulations (SGS) were computed. The Mean Image and the Standard Deviation maps were obtained, allowing the definition of High/Low risk clusters (Local G clustering) and the computation of spatial uncertainty. High-risk clusters are mainly distributed within the area with the highest altitude (agriculture/livestock) showing an associated low spatial uncertainty, clearly indicating the need for remediation. Atmospheric emissions, mainly derived from the metallurgical industry, contribute to soil contamination by PTEs.

(Research work is published in <https://doi.org/10.1016/j.scitotenv.2017.06.068>)

**Key words:** PTEs contamination, Bayesian Networks, Bayesian Risk Index, Sequential Gaussian Simulation, Local G clustering