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Daily Concentrations of PM_{2.5} in the Valencian Community Using Random Forest for the Period 2008–2018 [†]

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Abstract: Fine particulate matter (PM_{2.5}) is a global problem that affects the population health and contributes to climate change. Remote sensing provides useful information for the development of air quality models. This work aims to obtain a daily model of PM_{2.5} levels in the Valencian Community with a resolution of 1 km for the period 2008–2018. MODIS-MAIAC images, meteorological parameters of the MERRA-2 project, land cover information and ground level measurements of PM_{2.5} levels were analysed with Random Forest. The verification of the model was carried out using cross-validation repeated ten times, and an evaluation of a test set with 20% of the collected information. The final model was used to generate maps of the daily concentrations of PM_{2.5} for the area of the Valencian Community throughout the study period.

Keywords: PM2.5; LUR; Random Forest; MODIS; MERRA-2

1. Introduction

Atmospheric aerosols are a critical compound in the atmosphere due to their influence on climate change and population health [1]. Remote sensing has contributed significantly to the air quality study in order to capture the spatial-temporal variation of pollutants. Previous papers have shown that the amount of light absorbed or scattered by suspended particles, aerosol optical depth (AOD), is a relevant parameter for estimating $PM_{2.5}$ at ground level [2,3].

Moderate Resolution Imaging Spectroradiometer (MODIS) products are widely used in atmospheric models, as they have a daily review and a convenient spatial resolution for regional and local studies [2]. The recent Multi-angle Implementation of Atmospheric Correction (MAIAC) algorithm presents new opportunities for the development of atmospheric aerosol models [3]. These images employ the MODIS Aqua and Terra data and improve spatial resolution from 25 to 1 km [3].

Land use regression (LUR) models have had a broad application in different parts of the world. The LURs incorporate satellite images and meteorological and land use information as predictors to model PM_{10} (particulate matter with diameter < 10 μ m) and $PM_{2.5}$ (particulate matter with diameter < 2.5 μ m). Recent work shows positive results in the use of Random Forest (RF) for LUR models [2].

The air of the Valencian Community is affected by the growth of the vehicle fleet, industrial production, Sahara dust events and biomass burning smoke. An air quality model would allow the risk of exposure to be estimated. Previous works presented the correlation between daily PM_{2.5} ground-measures and AOD MODIS for this region [4,5]. The aim of this work is to apply an RF model, using 15 atmospheric

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variables and characteristic for land use to estimate the daily PM_{2.5} ground-concentration at the 1 km grid for 2008–2018 in Valencia Community.

2. Materials and Methods

2.1. Study Area

The Valencian Community is to the east of the Iberian Peninsula, on the Mediterranean coast (Figure 1). The population is mainly concentrated in urban centres, in particular, in the metropolitan areas of Valencia and Alicante. Segura et al. [6] found a significant relationship between black smoke and the number of emergency admissions for heart disease in Valencia city.

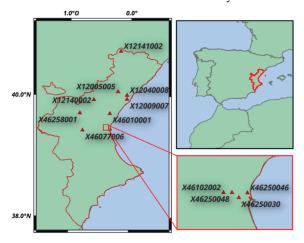


Figure 1. Valencian Community and the PM2.5 monitoring stations.

2.2. Data Sets

The hourly concentrations at the ground level of PM_{2.5} were download from the *Valencian Network for Monitoring and Control of Atmospheric Pollution of the Generalitat Valenciana* for the period between 1 January 2008 to 30 September 2018 (http://www.agroambient.gva.es/va/web/calidad-ambiental). During this time, 24 stations measured PM_{2.5} continuously, of which only 12 stations had a percentage of missing values less than 30% (Figure 1). For this work, we calculate the average PM_{2.5} concentrations between the hours of the Aqua and Terra satellite overpass.

MODIS-MAIAC products were downloaded from the *Level-1* and *Atmosphere Archive* and *Distribution System* website (https://ladsweb.modaps.eosdis.nasa.gov/) [2]. AOD measurements were calibrated with Aerosol Robotic Network (AERONET) data Level 2.0 (http://aeronet.gsfc.nasa.gov/) [4]. The fraction of the artificial surface was estimated for each pixel using the information provided by the Corine Land Cover project for the year 2012 [7]. The terrain elevation was obtained from the *Consultative Group for International Agricultural Research* Consortium for Spatial Information GEOPortal (http://srtm.csi.cgiar.org) with a resolution of 90 m at the equator [8].

Finally, atmospheric conditions data was download from the *NASA's Goddard Earth Sciences Data and Information Services Center* website (https://disc.gsfc.nasa.gov/). The Modern Era-Retrospective Analysis for Research and Applications, Version 2 (MERRA-2) is a global atmospheric reanalysis using the Goddard Earth Observing System Model, Version 5 with its Atmospheric Data Assimilation System, at a spatial resolution of $0.5^{\circ} \times 0.625^{\circ}$ [9].

2.3. Statistical Analysis

The RF model was trained with 80% of the collected information and evaluated with the remaining 20%. The model was built using PM_{2.5} observations as dependent variables. The predictor variables of the model were: (1) atmospheric variables: aerosol optical depth (AOD), surface pressure (PS), relative humidity (RH), surface temperature (T), surface wind component u (U), surface wind component v (V), black carbon surface mass concentration (BCSMASS), dimethyl-sulfide surface mass concentration

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(DMSSMASS), dust surface mass concentration (DUSMASS), SO₄ surface mass concentration (SO4SMASS), sea salt surface mass concentration (SSSMASS25), total precipitation (PRECTOT), high cloud cover (CLDHGH), low cloud cover (CLDLOW); (2) land use: fraction of artificial surface (CLC_1); (3) terrain elevation (DEM). The data were centered and scaled prior to being incorporated into the model.

The model verification contains a 10-fold cross validation (cv). The feature importance of each variable is then calculated after the model fitting process. Analyses were performed using the R language [10] and the "caret" package for RF model [11]. The maps were made with the software QGIS [12].

3. Results and Discussion

The ground $PM_{2.5}$ average for the entire period was 8.3 $\mu g.m^{-3}$. The months with the highest concentration were February (10.9 $\mu g.m^{-3}$) and March (10.4 $\mu g.m^{-3}$) and the lowest May (6.13 $\mu g.m^{-3}$) and June (7.22 $\mu g.m^{-3}$). Station X46250048 was the site with the most elevated average $PM_{2.5}$ levels (11.8 $\mu g.m^{-3}$). This station is in a busy urban area of the city of Valencia. The lowest station (X12141002, 6.13 $\mu g.m^{-3}$) is situated in Viver, a small village of fewer than 2000 inhabitants (Figure 1).

The top RF accuracy with a ntree = 10 was with mtry = 7. The variables PRECTOT, CLDHGH and AOD were the most significant predictors that contribute to the model construction process. PRECTOT and CLDHGH have a pronounced influence on the deposition and dispersion of pollutants. Based on these results, we ran the daily model for the rest of the Community of Valencia. Figure 2a presents a map example of average PM₂₅ concentrations for the years 2008.

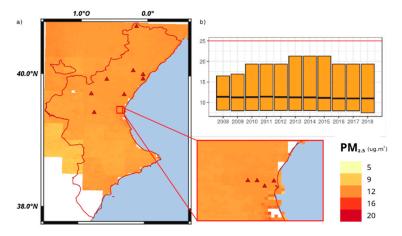


Figure 2. (a) Example of average PM₂₅ concentrations map in 2008 for the Valencian Community and the city of Valencia. (b) Range of annual mean concentrations modelled for the Valencia Community (2008–2018).

Results indicate a high predictive capability of the RF model, with an extremely high R^2 (0.89). The predictions for the test set have a good fit, with a root-mean-square error (RMSE) equal to 2.29 and a mean absolute error equal (MAE) to 0.67. In turn, the errors in estimations occur when modelling the highest concentrations. These may be a consequence of two factors. On the one hand, high values are underrepresented in the data set because high concentrations are particular events in the time series. On the other hand, the elevated measurements could be due to situations strongly influenced by events of micro-scale or of short temporal duration. Problems of immeasurability appear when modelling a point data from variables registered in portions of area (pixel).

The European Directive 2008/50/EC fixes the annual concentration for PM_{2.5} in 25 μ g.m⁻³. The model does not show areas that exceeded this annual level (Figure 2b). The year with the smallest spatial variation was 2008, with a minimum annual modelled concentration of 8 μ g.m⁻³ and a maximum of 16 μ g.m⁻³, while the year with the most considerable variation was 2015 (range: 7–21 μ g.m⁻³).

4. Conclusions

This study proposed a daily concentration model of PM_{2.5} based on the RF for the Valencia Community (Spain). The method used AOD MAIAC measures, MERRA-2 products and land cover information to

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simulate ground PM_{2.5} values. Based on the evaluation of the 10-fold cross-validation method and the test set verification method, the model performs very well. With this data, we were able to predict ~90% of the temporal and spatial variability of PM_{2.5}. More RF trees and an exhaustive analysis of predictor variables will bring benefits to PM_{2.5} simulations in the future. This work may provide support for air quality management and may also give evidence for epidemiological studies.

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Conflicts of Interest: Authors declare no conflict of interest.

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