



Agroclimatic and spectral regionalization for soybean in different agricultural settings in the state of Paraná, Brazil

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Abstract

Information related to the climate, sowing time, harvest, and crop development is essential for defining appropriate strategies for agricultural activities, which helps both producers and responsible bodies. Paraná, the second largest soybean producer in Brazil, has high climatic variability, which greatly influences planting, harvesting, and crop productivity periods. Therefore, the objective of this study was to regionalize the state of Paraná, considering decennial metrics associated with climate variables and the enhanced vegetation index (EVI) during the soybean cycle. Individual and global analyses of these metrics were conducted performed using multivariate techniques. These analyses were carried out in agricultural scenarios with low, medium, and high precipitation, corresponding to harvest years 2011/2012, 2013/2014, and 2015/2016, respectively. The results obtained from the scores of the retained factors and the cluster analysis were the profile of the groups, with Group 1 presenting more favourable climatic and agronomic conditions for the development of soybean crops for the three harvest years. The opposite occurred for Groups 2 (2011/2012 and 2013/2014) and Group 3 (2015/2016). During the soybean reproductive phases (R2 – R5), precipitation values were inadequate, especially for Group 2 (2011/2012 and 2013/2014) with high water deficit, resulting in a drop in soybean productivity. The climatic and agronomic regionalization of Paraná made it possible to identify the regions most suitable for growing soybeans, the effect of climatic conditions on phenological stages, and the variability of soybean productivity in the three harvest years.

Introduction

In Brazil, soybean stands out as one of the main commodities of major relevance to the country's economy, showing high production and leading exports. The soybean production in Brazil was 135.4 million tons in the 2020/2021 harvest year, reinforcing Brazil's leading position in the international market (USDA 2021).

Paraná stands out among the soybean-producing states, being the second largest producer in Brazil and responsible for 14.6% of the production in the 2020/2021 harvest year. However, the state's production showed a 5% reduction in relation to the 2019/2020 harvest year, mainly owing to weather conditions (CONAB 2021).

Climatic factors directly influence the development, quality, and productivity of agricultural crops (Stansel and Fries, 1980; Baldisera and Dallacort, 2017). In this sense, knowledge of meteorological conditions and productivity trends helps agribusiness companies and official bodies identify and manage the risks associated with extreme weather events, such as droughts and floods.

With accurate information about the cultivated area and expected productivity, agribusiness companies and official bodies can make more assertive decisions regarding when and how to market their crops. Furthermore, this information is relevant for monitoring agricultural production, predicting food supply, and taking preventive measures in the event of shortages or food crises (MAPA 2022). Producers also need this information so that they can, based on stocks and expected future prices, make decisions regarding the crop to be cultivated and also choose the best sowing time, aiming for greater productivity.

Therefore, the sowing time for annual crops depends on the occurrence of precipitation and temperature conditions, which may differ between regions (Oliveira and Padovani, 2017). Rainfall is the main source of water for plants and affects practically all physiological and biochemical processes. The amount and distribution of precipitation affects the availability of water in the soil, which is essential for plant growth. Lack of rain can lead to drought and water stress, whereas excess rain can cause waterlogging and drainage problems (Farias *et al.*, 2007; Cohen *et al.*, 2021).

Air temperature regulates plant growth and development rates (Serio *et al.*, 2006). Furthermore, other variables such as solar radiation and potential evapotranspiration influence agricultural production. Solar radiation provides the energy necessary for photosynthesis, which is essential for plant biomass production. The amount of sunlight directly affects the growth rate and yield of crops (Taiz and Zeiger, 2009).

The relationship between climate variables and agriculture is fundamental to the success of a production system (Baldisera and Dallacort, 2017). The climate directly influences the growth, development, and productivity of crops. In this way, strategies such as carrying out agro-climatic zoning make it possible to understand the climatic characteristics of a given region and determine the best agricultural practices and crops that are most suitable for that environment (Ribeiro *et al.*, 2015).

However, in places with continental characteristics, such as Brazil, on-site monitoring of these variables is a difficult and costly task to carry out. Therefore, agricultural monitoring with remote sensing allows monitoring of the crop throughout its cycle, ultimately generating a more objective estimate of the agricultural harvest (Johann *et al.*, 2016; Becker *et al.*, 2017; Paludo *et al.*, 2020). Using remote sensing techniques, Becker *et al.* (2020) identified phenological patterns with periods of growth and senescence based on vegetation index data for agricultural crops from the Moderate Resolution Imaging Spectroradiometer (MODIS) sensor. Using these techniques, the authors monitored the vegetation phenology quickly and on a large scale.

Furthermore, remote sensing allows the acquisition of a vast amount of agrometeorological and crop development data, both spatially, and temporally, often generating Big Data (Corea, 2019). In this context, we sought to understand the spatial and temporal variability of climatic variables and the crop vegetation index, both at the global and regional levels, using multivariate analysis techniques, such as exploratory factor analysis (EFA) and clustering.

Cluster analysis techniques were used by Johann *et al.* (2013), who identified five homogeneous regions that characterized the temporal dynamics of the normalized difference vegetation index (NDVI) obtained by the Satellite Pour l'Observation de la Terre (SPOT) satellite for soybean-producing municipalities in Paraná.

This technique was also used by Shirin and Thomas (2016) and Sathiaraj *et al.* (2018), who collected daily precipitation and temperature information to characterize homogeneous regions in India and the United States.

Other studies used EFA and clustering techniques to characterize an agricultural region, such as Bianchi *et al.* (2016), who evaluated the production of sugarcane, corn, cassava, and soybeans in the middle Paranapanema region, in the state of São Paulo, and Grzegozewski *et al.* (2020), who evaluated three consecutive harvest years of soybean cultivation, in relation to the variables precipitation, potential evapotranspiration, solar radiation, minimum, average, and maximum temperatures, improved vegetation index (EVI), and soybean productivity to identify five homogeneous regions, for the municipalities of the state of Paraná.

The aim of this study is to evaluate homogeneous regions for soybean cultivation in Paraná. Previous studies, such as those by Lopes *et al.* (2019), Gebert *et al.* (2018), Richetti *et al.* (2018), and Nascimento *et al.* (2020), used few climatic variables and did not consider the phenological stages of soybeans. The integration of meteorological variables and the vegetation index (EVI) with the phenological stages, in 25 km data in Paraná, has not yet been done. To fill these gaps, three agricultural scenarios were

evaluated to regionalize the state, considering decadal metrics and the European Centre for Medium-Range Weather Forecasts (ECMWF) grid, aiming to define an agroclimatic regionalization for soybeans, the state's main crop.

Materials and methods

Data and metrics

The study area encompasses the state of Paraná, Brazil, limited to 22°29'S and 26°43'S, and 48°2'W and 54°38'W, respectively. The state is subdivided into ten mesoregions (representing territorial units with homogeneous physical, economic, and social characteristics), totalling 399 municipalities (IBGE, 2018) (Fig. 1(a)).

To carry out this study, data from the enhanced vegetation index (EVI), meteorological variables, such as rainfall (mm) were used; minimum, average and maximum air temperature (°C); solar radiation (MJ/m²/day); potential evapotranspiration (mm) by the Penman–Monteith method; and climatological water balance (mm), proposed by Thornthwaite and Mather (1955), was calculated by taking the difference between precipitation and potential evapotranspiration (Pereira *et al.*, 2002), and the weighted average productivity of soybeans.

These meteorological variables were selected because they are critical for agricultural production and directly influence final productivity (Fontana *et al.*, 2001; Krüger *et al.*, 2007).

Based on the precipitation information from the ECMWF, three crop years (2011/2012, 2013/2014, and 2015/2016) were evaluated, representing different agrometeorological scenarios within a historical series from 2000/2001 to 2015/2016.

The 2011/2012 crop year showed typical drought behaviour, with the lowest average precipitation values. The 2015/2016 crop year, on the other hand, was characterized as rainy, with the highest average precipitation values. The 2013/2014 crop year had a precipitation close to the average of the historical series evaluated. As it was not possible to explore the entire time series, we selected crop years with these distinct characteristics to understand the productivity patterns under these conditions.

Furthermore, precipitation, which presents great spatial and temporal variability, is one of the most important climatic elements in agriculture because of its significant influence on all stages of plant development. Generally, it is the variable that most affects the final productivity of soybeans (Neumaier *et al.*, 2021). Using data from the EVI, meteorological variables, and estimates of sowing dates, maximum vegetative development and harvest, associated metrics were generated at 10-year intervals, considering the entire soybean crop cycle in Paraná (Figs 2 and 3(a)).

Enhanced vegetation index – EVI metrics

For the EVI, 14 metrics were generated associated with the 10-year period of the soybean cycle for each crop year analysed (Figs 2, 3(a, b)). To determine the metrics, EVI images from the MOD13Q1 and MYD13Q1 products from the MODIS sensor were initially used, with a spatial resolution of 250 m and a temporal resolution of 8 days (Fig. 1(d) and (e)) (NASA, 2019).

The spectro-temporal profile of the EVI was analysed for the harvest years 2011/2012, 2013/2014 and 2015/2016, using soybean mapping carried out by Souza *et al.* (2015), Grzegozewski *et al.* (2016), and Verica (2018), respectively. EVI values were extracted from the pixels of these mappings between scenes

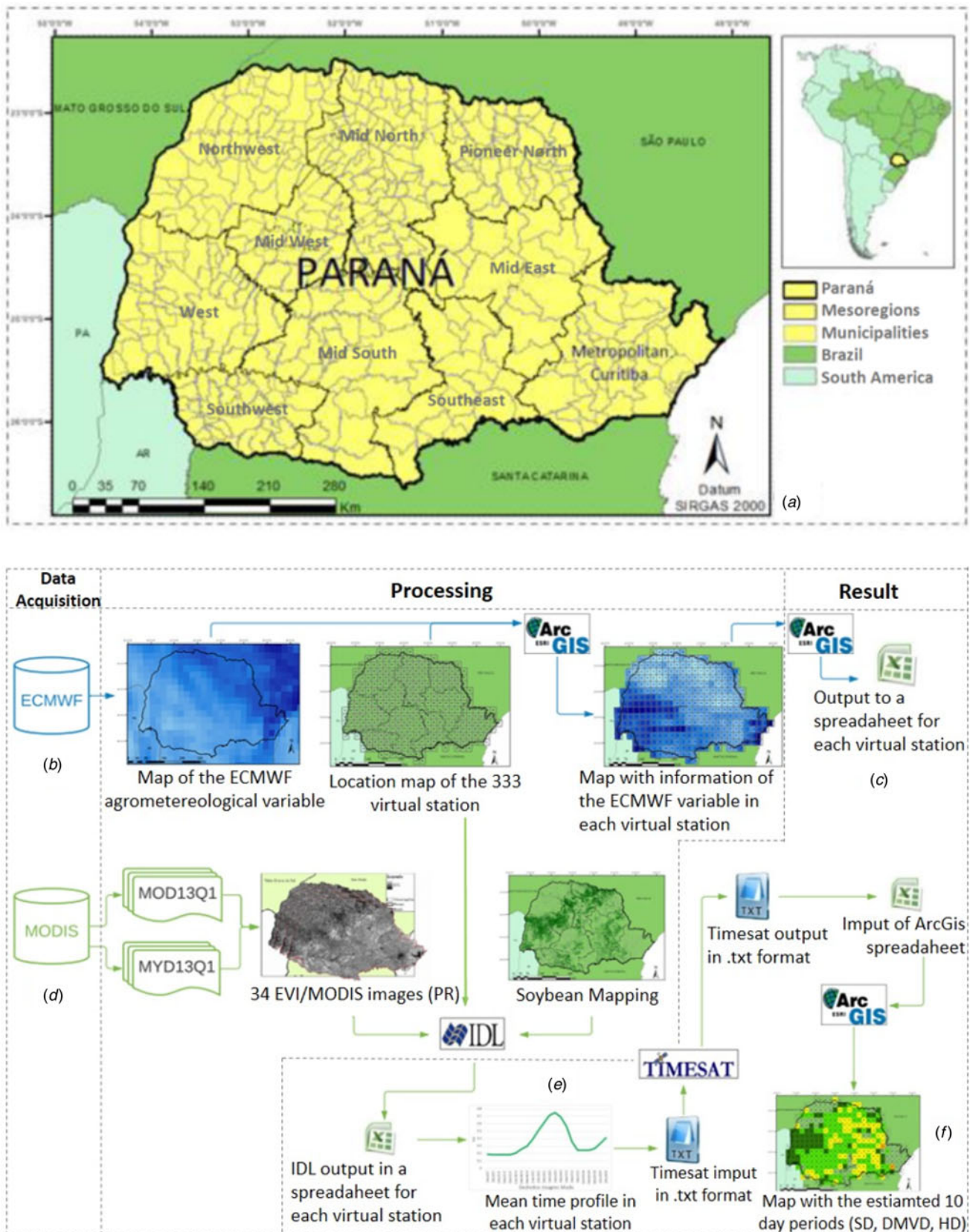
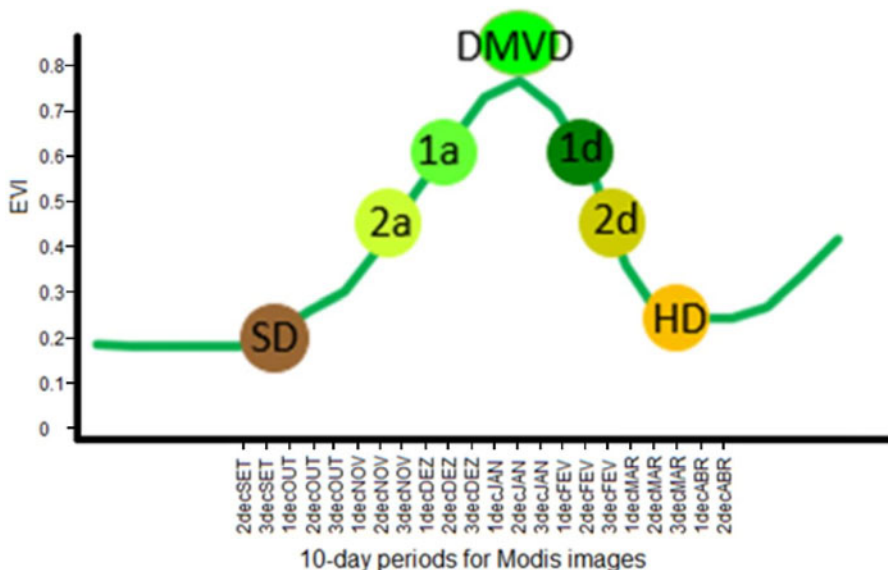


Figure 1. Flowchart representing: (a) Study area, state of Paraná. (b) Summary of methodological steps for data acquisition: European Centre for Medium-Range Weather Forecasts (ECMWF). (c) Various information on meteorological variables at each virtual station. (d) Data acquisition: Moderate Resolution Imaging Spectroradiometer (MODIS). (e) Generation of the average enhanced vegetation index (EVI) profile for each VS. (f) sowing date (SD), date of maximum vegetative development (DMVD) and harvest date (HD) estimates for each virtual station.

Figure 2. Representation of the soybean crop cycle, with sowing date, maximum vegetative development and harvest, and the 10-day periods close to date of maximum vegetative development. **EVI**, enhanced vegetation index; **SD**, 10-day sowing period; **2a**, Two 10-day periods before DMVD; **1a**, One 10-day period before DMVD; **DMVD**, 10-day period of maximum vegetative development; **1d**, One 10-day period after DMVD; **2d**, Two 10-day periods after DMVD; **HD**, 10-day harvest period; **2decSET**, **3decSET**: second and third 10-day periods in September. **1decOUT**, **2decOUT**, **3decOUT**: first, second and third 10-day periods in October. **1decNOV**, **2decNOV**, **3decNOV**: first, second and third 10-day periods in November. **1decDEZ**, **2decDEZ** **3decDEZ**: first, second and third 10-day periods in December. **1decJAN**, **2decJAN**, **3decJAN**: first, second and third 10-day periods in January. **1decFEV**, **2decFEV**, **3decFEV**: first, second and third 10-day periods in February. **1decMAR**, **2decMAR**, **3decMAR**: first, second and third 10-day periods in March. **1decABR**, **2decABR**, **3decABR**: first, second and third 10-day periods in April.



225(13/08) and 120(30/04) using the tenth day corresponding to the date of the scene for this study. The IDL (Interactive Data Language) programming language was used (Esquerdo et al., 2011).

With the EVI values on the decennial scale, ArcGis 10.3 software (ESRI, 2018) was used to obtain the EVI spectral profile at each virtual station (Fig. 1(e)). These profiles were analysed using TIMESAT software (Eklundh and Jönsson, 2015), through

the TSM_GUI module, and using the Savitzky–Golay filter (Savitzky and Golay, 1964). This allowed the analysis of temporal vegetation index profiles, generating data on the sowing date (SD), date of maximum vegetative development (DMVD), harvest date (HD), and crop cycle duration (CC) (Johann et al., 2016; Becker et al., 2020).

For each virtual season and harvest year, the sowing date, date of maximum vegetative development and harvest date were

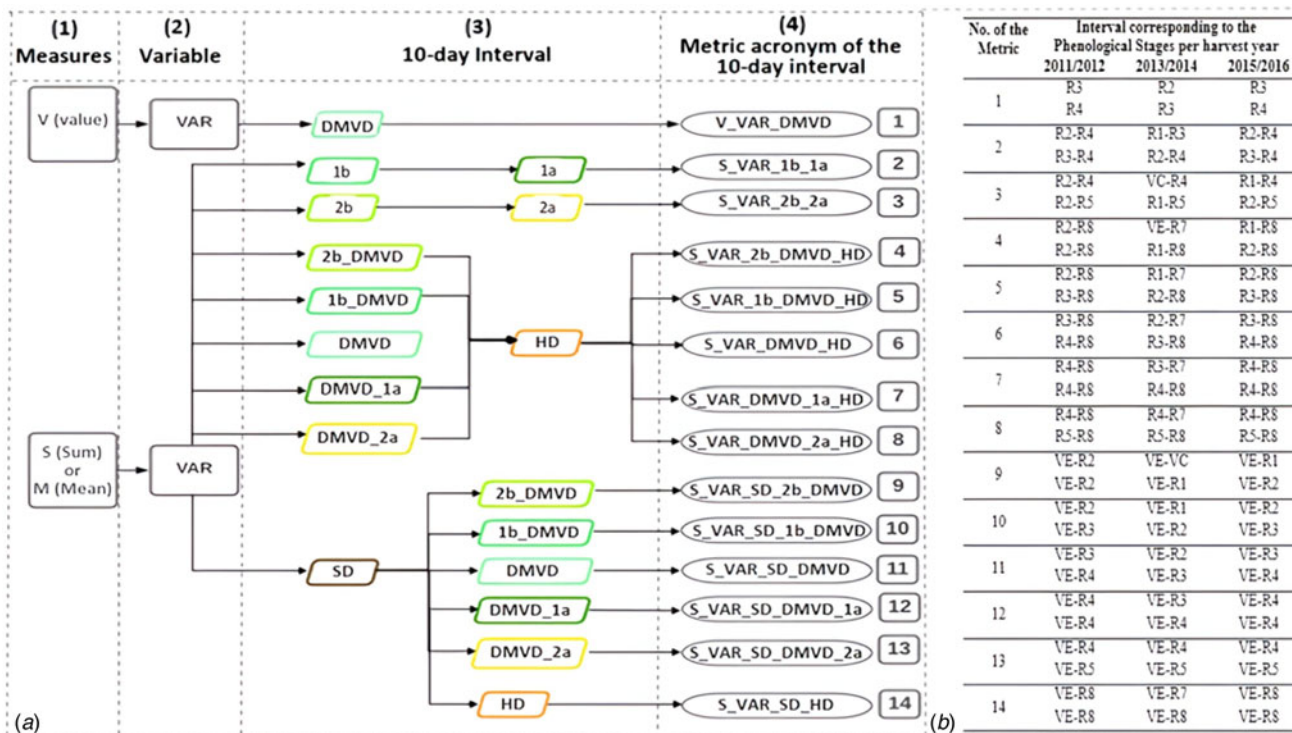


Figure 3. (a) Description of the variables and the acronyms that represent them, associated with 10 years, generated from the sowing date, date of maximum vegetative development and harvest date. SD: Sowing 10-day period, DMVD: 10-day period of maximum vegetative development, HD: Harvest 10-day period. (b) Metrics of meteorological variables and EVI associated with 10 years with the respective phenological stages of soybean, in each crop year. V: Value; S: Sum; VE: Emergence; VC: Cotyledon; R1: Beginning of flowering. R2: Full flowering. R3: Beginning of legume formation. R4: Full pod formation. R5: Grain filling. R6: Full grain. R7: Beginning of maturation. R8: Full maturation.

identified, which were rescheduled to 10 days, that is, according to the date found, they were assigned to the 1st, 2nd or 3rd tenth of the month (Fig. 1(f)).

During the process of estimating these data, the virtual stations were evaluated for the EVI spectral profile. Some of these virtual stations showed deviations such as high variability during the cultivation period, regions with low crop cultivation, and edge regions. The profiles that did not present a spectro-temporal pattern of EVI characteristic of soybeans were eliminated from the data set, leaving, respectively, for 2011/2012, 2013/2014 and 2015/2016, a number of 251, 245 and 268 virtual stations.

Weather metrics

For each of the meteorological variables, precipitation, potential evapotranspiration, water balance, solar radiation, and minimum, average, and maximum temperatures, 14 metrics were generated and associated with the 10-year period of the soybean cycle for each crop year analysed (Figs 2, 3(a, b)).

Meteorological data were obtained from the ECMWF model (ECMWF 2012) at a regular spatial resolution of 0.25 ° (± 25 × 25 km) longitude and latitude. For each point on this grid, called virtual station, information was obtained on the following variables: rainfall; minimum, average and maximum air temperature; Solar radiation; potential evapotranspiration by the Penman–Monteith method; and climatological water balance, proposed by Thornthwaite and Mather (1955), was calculated by taking the difference between precipitation and potential evapotranspiration (Pereira *et al.*, 2002). These variables were made available hourly to obtain the daily average for each variable.

Thus, 14 metrics were generated for the EVI and 98 for the meteorological variables, totalling 112 metrics associated with the 10-year period of the soybean cycle.

Metrics associate with soybean phenological stages

Variables associated with the phenological stages of soybean crops were generated. To prepare them and estimate the number of days and 10 days per phenological stage, the works of Fehr and Caviness (1977), Doorenbos and Kassam (1979), Kaster and Farias (2011), and Thomas (2018) were used (Table 1).

In this way, metrics associated with the phenological stages and the intervals between them were obtained, considering the accumulated sum of the EVI and the meteorological variables (precipitation, potential evapotranspiration, solar radiation, and water balance), in addition to the average for the meteorological temperature variables (minimum, average, and maximum), as shown in Fig. 3(a).

Thus, 29 metrics were generated for the accumulated sum and 29 metrics for the accumulated average associated with the

phenological stages and the intervals between them, totalling 232 metrics. For each crop year, the database was composed of 112 metrics associated with 10-year periods and 232 metrics associated with phenological stages, totalling 344 metrics.

For the two groups of metrics - associated with decennials and phenological stages - a preliminary analysis was carried out to calculate the canonical correlation coefficient between these two groups to assess the existence of an association between the metrics generated by decennials and stages, phenological. The canonical correlations were greater than 0.90 for the three crop years, indicating the variables generated for the two groups are similar.

Therefore, we chose to use only the metrics associated with 10 years (Fig. 3(a)), as they are simpler to obtain in future applications of this methodology. However, in the discussions, the metrics associated with 10-day periods were related to the metrics associated with phenological stages (Fig. 3(b)) for a better agroeconomic understanding.

It was found that there is high variability in phenological stages in relation to 10-year intervals throughout the soybean cycle in Paraná. For example, in the 2011/2012 and 2015/2016 harvest years, on the date of maximum vegetative development, soybeans were between the beginning of legume formation and full pod formation stages (R3 and R4, Fig. 3(b)), whereas that in 2013/2014, soybeans were in the full flowering and beginning of legume formation stages (R2 and R3, Fig. 3(b)). As a result, the phenological stages of soybeans were independently characterized for each crop year (Fig. 3(b)).

Thus, if the soybeans were in the R3 stage in the virtual station, for example, the 10-day interval, 1a_R3_1d, (one 10-day before the R3 stage and one 10-day after) was described, which corresponds to the R2 to R4 stages, and so on for the next 10-day intervals, until completing the soybean cycle with the corresponding phenological stages.

Exploratory factor analysis and dimensionality reduction

Considering the large number of metrics associated with EVI and meteorological variables for each crop year under study, an EFA was performed (Fig. 4) to reduce the dimensionality and evaluate the structure of the metrics through their correlations with the factors (Hair *et al.*, 2014).

After organizing and reading the data, a correlation matrix was obtained using Pearson’s linear correlation. The adequacy of the correlation matrix was verified using Bartlett’s Test of Sphericity and the KMO Index (Hair *et al.*, 2014). Eigen extraction was performed using principal component analysis, and the retention of factors in the EFA followed Kaiser’s (1958) criteria (eigenvalues $\lambda > 1$) in addition to considering a minimum percentage of 70% of the total variability of the metrics associated with the EVI and

Table 1. Number of days and ten-day periods (in between parenthesis) estimated by phenological stage based on the 120-day cycle proportion

SD-VE	VC-V(n)	R1	R2	R3	R4	R5	R6	R7	R8	Total Cycle
9 (1)	34 (3)	10 (1)	5 (1)	12 (1)	9 (1)	12 (1)	9 (1)	12 (1)	8 (1)	120 (12)
10 (1)	37 (4)	11 (1)	5 (1)	13 (1)	10 (1)	13 (1)	9 (1)	13 (1)	9 (1)	130 (13)
10 (1)	40 (4)	11 (1)	7 (1)	13 (2)	11 (1)	16 (1)	8 (1)	14 (1)	10 (1)	140 (14)
11 (1)	42 (4)	13 (1)	6 (1)	16 (2)	10 (1)	18 (2)	9 (1)	14 (1)	11 (1)	150 (15)

SD, 10-day sowing period; VE, Emergence; V; Cotyledon; V1, First node; V(n), nth node; R1: Beginning of flowering. R2: Full flowering. R3: Beginning of legume formation. R4: Full pod formation. R5: Grain filling. R6: Full grain. R7: Beginning of maturation. R8: Full maturation.

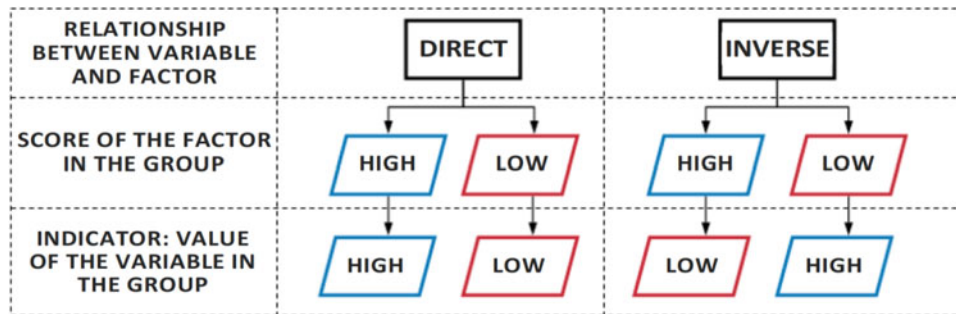


Figure 4. Flowchart corresponding to the analysis of the profile of virtual stations within each group. The relationships between the decennial metrics associated with the enhanced vegetation index (EVI) and the meteorological variables and factors (direct (positive) or inverse (negative)). The value of the metrics in the group is called an indicator (high (positive) or low (negative)); these indicators are high or low in relation to the doctor in the group.

meteorological variables explained by the factors (Ferreira, 2018). The varimax orthogonal rotation method was used (Damásio, 2012; Hair et al., 2014). With the factors retained, their scores were estimated for each virtual station. Pearson’s linear correlation was verified between the variables and factors, with a correlation above 0.60. Furthermore, the direct or inverse relationship between variables and factors was evaluated (Fig. 5). With these scores, it was possible to perform cluster analysis (Fig. 4).

Grouping of meteorological zones and EVI

To obtain these groups, tests were carried out using hierarchical methods and the non-hierarchical k-means method (MacQueen, 1967), using Euclidean distance as a measure of dissimilarity.

The hierarchical methods did not provide adequate values for the cophenetic correlation coefficient. Therefore, the non-hierarchical k-means method was used to form similar groups (Fig. 4).

The optimal number of groups was determined using the following cluster validation indices: Calinski–Harabasz (CH) (Calinski and Harabasz, 1974), Dunn (Dunn, 1974), silhouette index (ISil) (Rousseeuw, 1987), DB index (Davies and Bouldin, 1979), C index (Hubert and Levin, 1976), SD index (Halkidi et al., 2000), CCC (Sarle, 1983), Scott (Scott and Symons, 1971), and KL index (Krzanowski and Lai, 1988).

To evaluate the profile of the virtual stations belonging to each group, the relationships (direct or inverse) between the variables and factors as well as the score of each factor within each group (high or low) were verified. With this, the value of the variable

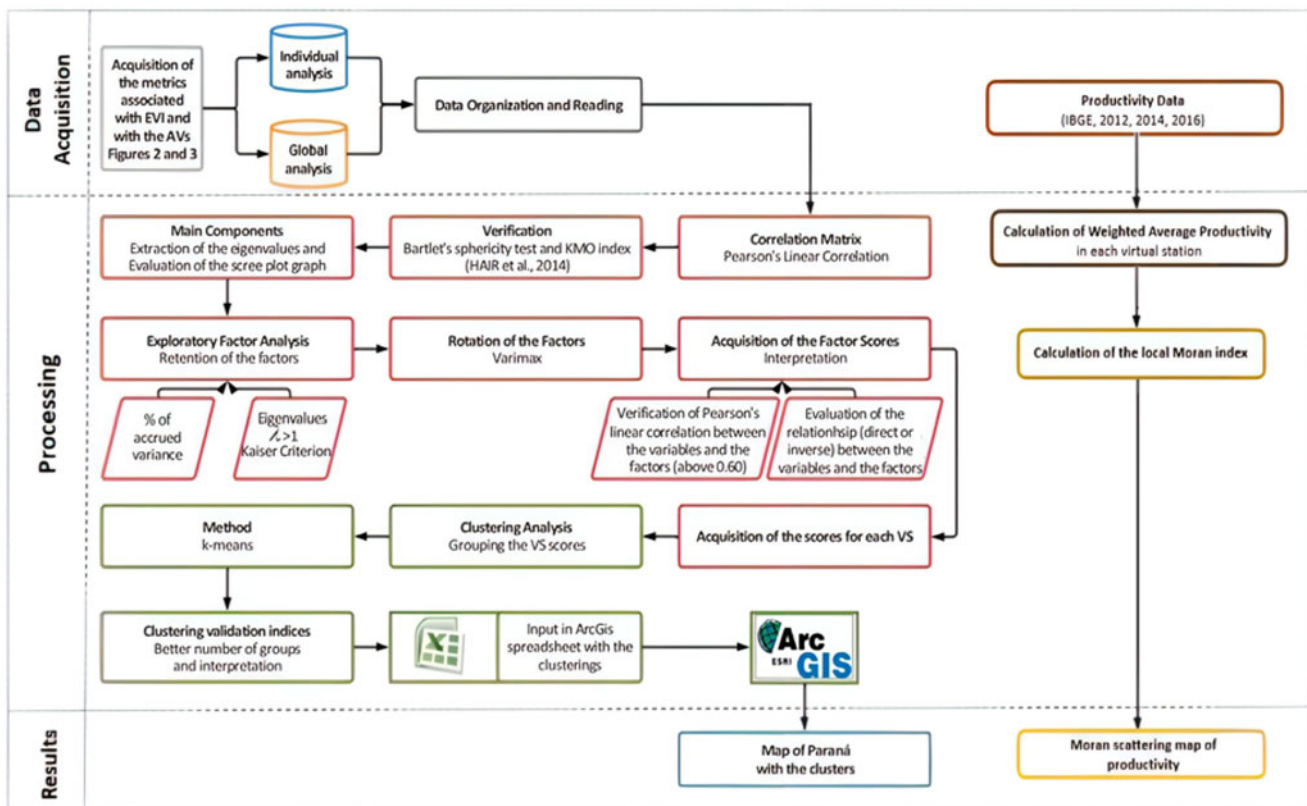


Figure 5. Flowchart corresponding to the stages of individual and global analysis of metrics associated with enhanced vegetation index (EVI) and meteorological variables, using multivariate methods. Steps in constructing the Moran scattering map from productivity data.

in the group, called an indicator (high or low), was obtained, as shown in Fig. 5.

Individual analysis

Fourteen metrics were used for individual analysis (Fig. 3(a)) associated with the EVI and meteorological variables. The EFA and grouping were carried out with the metrics associated with the EVI and the meteorological variables individually, resulting in maps with the groupings of metrics for each meteorological variable and the EVI in each crop year (Fig. 6).

Global analysis

For the global analysis, 14 metrics (Fig. 3(a)) associated with the EVI and meteorological variables were also used. The EFA and grouping were performed considering all metrics simultaneously (totalling 112 metrics analysed), resulting in a single map with the groupings of all metrics associated with the EVI and meteorological variables for each crop year (Fig. 7). The stages of these analyses are illustrated in Fig. 5.

Individual analysis allows the visualization and understanding of the spatial pattern of metrics associated with each of the meteorological and EVI variables. In global analysis, although it does not provide this specific visualization, it allows us to understand the pattern of metrics associated with the EVI and meteorological variables in a joint analysis.

Integration of soybean productivity data by municipality with the ECMWF grid

As the information on the EVI metrics and meteorological variables was integrated with the ECMWF grid on the Paraná map, it was also necessary to obtain information on the average soybean productivity in each ECMWF virtual station. To carry out this integration, it was initially verified which municipalities make up each virtual station and the percentage of areas of these municipalities in each virtual station.

Soybean productivity data (t/ha) were obtained from the 399 municipalities in the state of Paraná, from the Brazilian Institute of Geography and Statistics (IBGE) (IBGE, 2012, 2014, 2015) (Fig. 5). There were no average productivity data in 37, 28 and 23 municipalities for the harvest years 2011/2012, 2013/2014 and 2015/2016, respectively, due to the absence of soybean production.

For each virtual station, the Weighted Average Productivity (WAP) (Equation (1)) was calculated considering only the municipalities that present soybean production.

$$WAP = \frac{\sum_{i=1}^n a_i \cdot p_i}{\sum_{i=1}^n a_i}, \quad (1)$$

where WAP is the Weighted Average Productivity value, a_i is the percentage of the area of the i th municipality within the virtual station, p_i is the productivity of the i th municipality with $p_i \neq 0$ and $i = 1, \dots, n$ where n is the number of municipalities within the virtual station.

Spatial indicators

Moran's Local Index (Equation (2)) was used to identify virtual stations that were distinct in terms of productivity in relation to

other virtual stations.

$$I_i = \frac{n(x_i - \bar{x}) \sum_{j=1}^n w_{ij}(x_j - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}. \quad (2)$$

where n is the number of virtual station under study, x_i and x_j are the values of the variable of interest in virtual station i and j , with $i = 1, \dots, n$ and $j = 1, \dots, n$; \bar{x} is the average value of the variable of interest in the virtual station under study and w_{ij} are the elements of the spatial proximity matrix.

The local spatial association indicator (LISA) is commonly conceived as a numerical measure of a random variable that relates each individual data point to values observed in neighbouring locations (Anselin, 1995; Mathur, 2015).

All statistical analyses were performed using the R software version 4.0.2 (R Development Core Team, 2022). For EFA, the nFactors package was used (Raiche *et al.*, 2013), for validation measures, NbClust (Charrad, *et al.*, 2022), for the Moran Local index the rgeoda package (Li and Anselin, 2023) was used, and the maps were represented using ArcGIS software version 10.3 (ESRI, 2018).

Results

The results of the factor analysis and regionalization of the state of Paraná are presented through a cluster analysis.

Exploratory factor analysis

Overall, the variation explained by the three factors of the 2011/2012 harvest and by the four factors of the 2013/2014 and 2015/2016 harvests was greater than 82% for the global analysis and greater than 87% for the individual analysis, indicating that these factors define the dimensionality of the variables analysed and will be used in the subsequent steps (Table 2).

The first factor (F1) of the global analysis has more than 54% of the total variance of the data set, representing the water deficit during growth until grain development in the 2011/2012 harvest year. For the 2013/2014 crop year, this factor represents the conditions of interference of photosynthetic activity in plant development, accounting for more than 47% of the total variance. In the 2015/2016 harvest year, more than 43% of the total variance was represented by this factor, indicating a change in land cover owing to thermal conditions.

The second factor (F2) of the global analysis explains 21, 19, and 23% of the total variance for the harvest years 2011/2012, 2013/2014, and 2015/2016, respectively (Table 2), representing the influence of the variable climate on potential evapotranspiration during plant maturation. Factors 3 and 4 presented values below 13% of the total variance in the three harvest years (Table 2).

In the individual analysis, in which each variable was evaluated separately, the minimum, average, and maximum temperatures were retained in the first factor (F1), with a percentage of total variance greater than 82, 79, and 72%, respectively. Thus, F1 represents the information on these temperatures from emergence to soybean grain-filling for the three harvest years (Table 2).

It is also worth highlighting solar radiation, which was best explained by F1 in the low precipitation scenario (harvest year 2011/2012), representing 72% of the total variability of information during the grain-filling stages until maturation. This suggests that, when solar radiation reaches extreme intensities, plants can

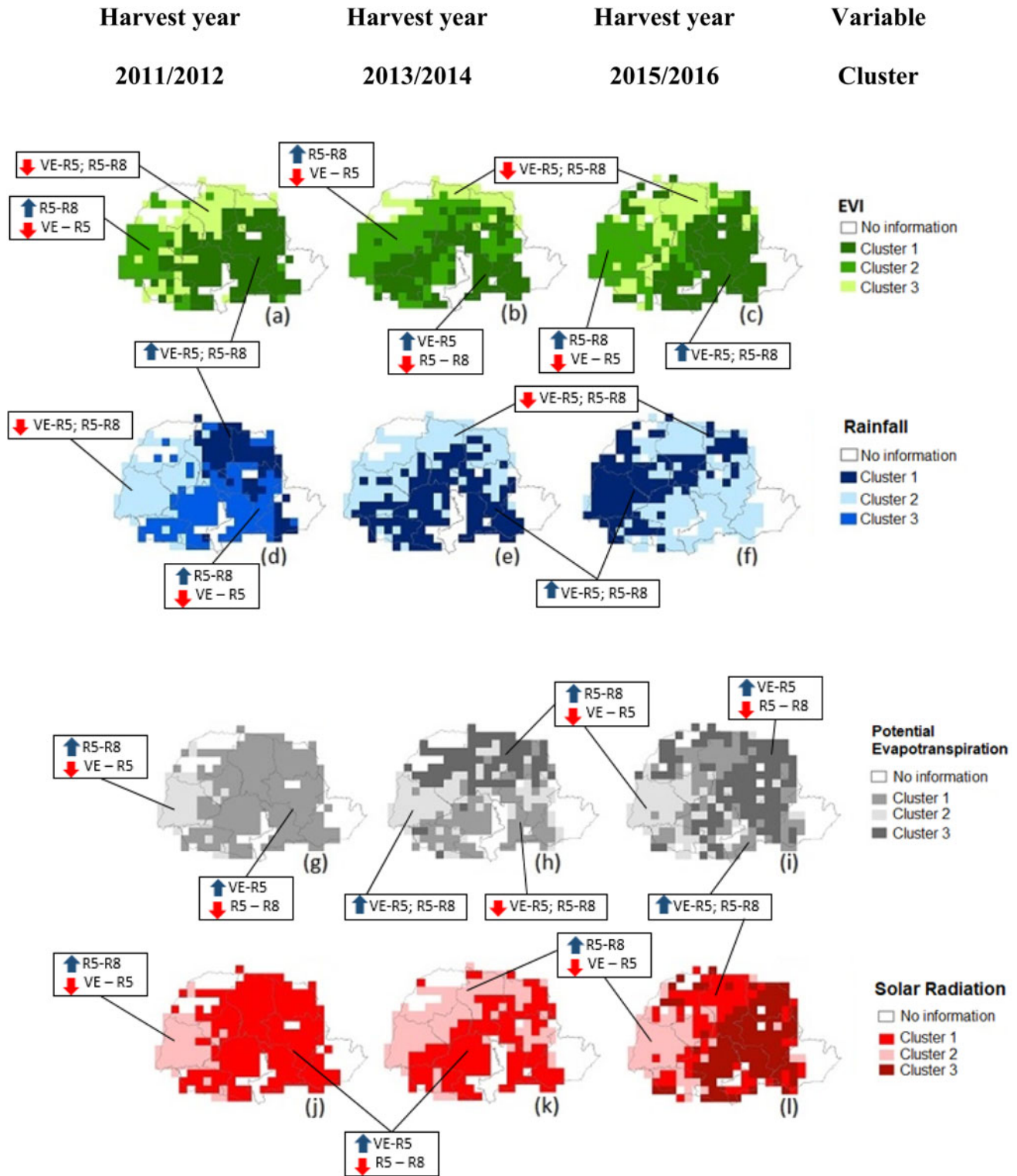


Figure 6. Regionalization of Paraná for individual analysis, considering the factorial scores of the virtual stations through cluster analysis, in relation to the metrics associated with the enhanced vegetation index (EVI) and indirect variations in each crop year, 2011/2012, 2013/2014, 2015/2016. (a) EVI (2011/2012), (b) EVI (2013/2014), (c) EVI (2015/2016), (d) Precipitation (2011/2012), (e) Precipitation (2013/2014), (f) Precipitation (2015/2016), (g) Potential Evapotranspiration (2011/2012), (h) Potential Evapotranspiration (2013/2014), (i) Potential Evapotranspiration (2015/2016), (j) Solar Radiation (2011/2012), (k) Solar Radiation (2013/2014), (l) Solar Radiation (2015/2016), (m) Water Balance (2011/2012), (n) Water Balance (2013/2014), (o) Water Balance (2015/2016), Temp: Temperature, (p) Minimum Temp (2011/2012), (q) Minimum Temp (2013/2014), (r) Minimum Temp (2015/2016), (s) Average Temp (2011/2012), (t) Average Temp (2013/2014), (u) Average Temp (2015/2016), (v) Maximum Temp (2011/2012), (w) Maximum Temp (2013/2014), (x) Maximum Temp (2015/2016). VE – R5, phases from sowing to grain filling; R2 – R5, phases from the beginning of flowering to grain filling; R5 – R8, stages from grain filling to maturation; DS – DC, 10 years from sowing to 10 years from harvest.

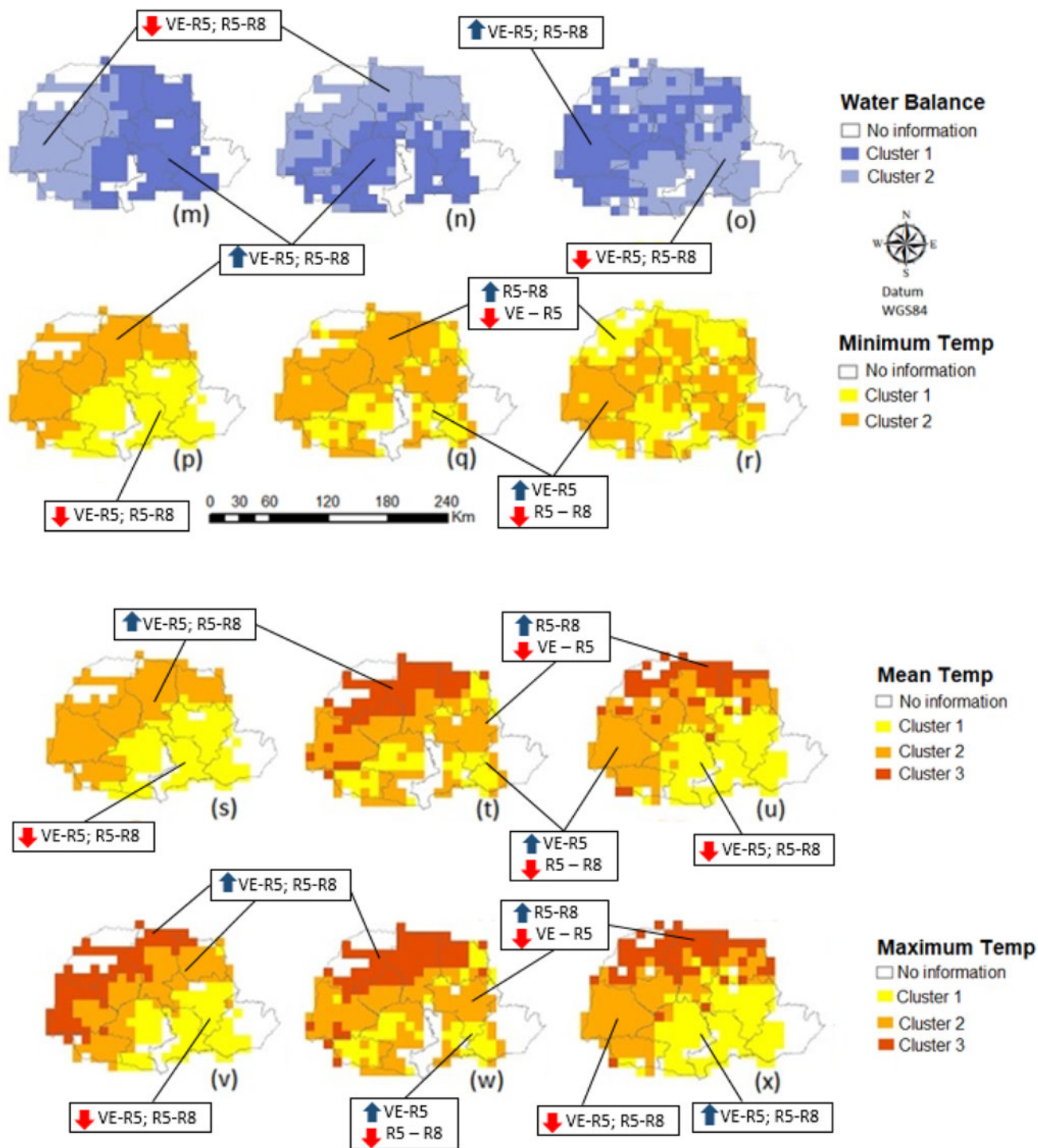


Figure 6. Continued.

tilt their leaves to reduce solar radiation interception, which reduces photosynthetic efficiency, especially if associated with low water availability (EMBRAPA, 2011), as occurred in this crop year.

Spatial regionalization of clusters

Considering both the individual analysis (Fig. 6) and the global analysis (Fig. 7), the profile of the clusters was created based on

the factors obtained in the factor analysis, together with the metrics associated with the EVI and meteorological variables, during the phenological stages of soybean cultivation. Figures 6 and 7 describe the phenological stages of each group, as detailed in Fig. 3(b), as well as the respective high and low indicators. The high and low indicators were compared between the harvest years 2011/2012, 2013/2014 and 2015/2016 and within each group. Still, these variables had consistently high or low values in each crop year and in which phenological stage these values were observed.

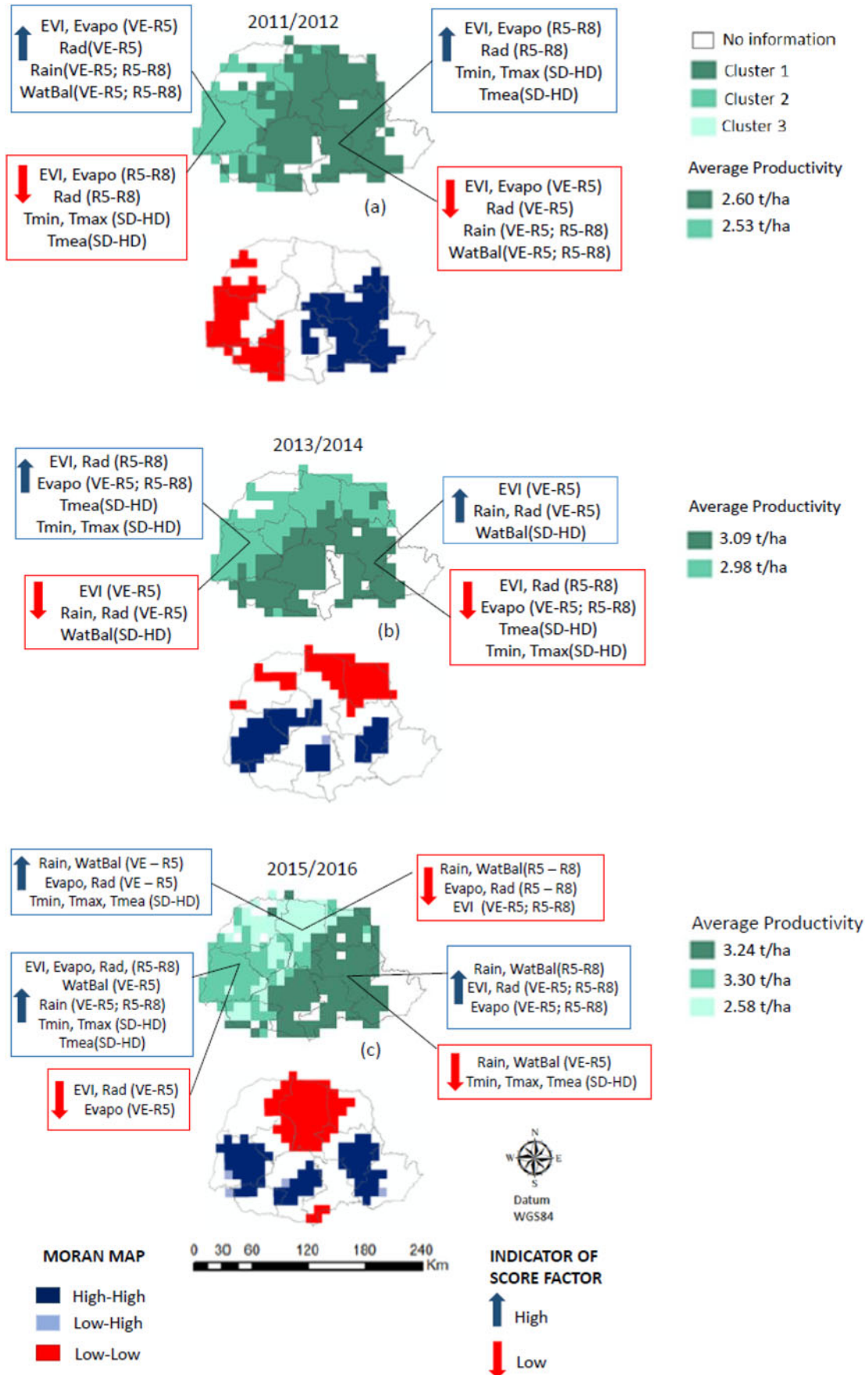


Figure 7. Regionalization of Paraná for global analysis, considering the factorial scores of the virtual stations through cluster analysis, in relation to the metrics associated with the EVI and meteorological variables in each crop year. Productivity information in each group of the respective harvest year. Moran scatter map for productivity for the respective crop year. (a) 2011/2012. (b) 2013/2014. (c) 2015/2016. VE – R5, phases from sowing to grain filling; R2 – R5, phases from the beginning of flowering to grain filling; R5 – R8, stages from grain filling to maturation; DS – DC, 10 years from sowing to 10 years from harvest (Fig. 3(b)).

Table 2. Number of factors and percentage of the total explained variance for the metrics associated with the EVI and the meteorological variables, resulting from the joint and individual analyses, for all three harvest years

Metrics associated with the EVI and the MVs	2013/2014										2015/2016									
	2011/2012					% of Explained variance					2015/2016									
	F1	F2	F3	F4	Accum.	F1	F2	F3	F4	Accum.	F1	F2	F3	F4	Accum.					
GA	54	21	9	84	47	19	10	6	82	43	23	13	9	88						
IA	50	37	8	95	56	22	13	92	92	52	32	11	95	95						
Evapo	55	38		93	47	37	11	95	95	48	38	9	95	95						
Rad	72	14	10	96	55	25	12	92	92	47	40	8	95	95						
WatBal	62	25		87	68	15	11	94	94	53	29	12	94	94						
EVI	52	35	12	99	45	40		85	85	57	32	10	99	99						
Tmin	90	4		94	82	13		95	95	95	2		97	97						
Tmea	89	5		94	79	14		93	93	95	2		97	97						
Tmax	85	6		91	72	19		91	91	91	5		96	96						

MVs, Meteorological Variables; GA, Global Analysis; IA, Individual Analysis; Evapo, Evapotranspiration; Rad, Radiation; WatBal, Water Balance; T, Temperature; min, Minimum; me, Mean; max, Maximum; F, Factors; Accum., Accumulated.

In the individual analysis (Fig. 6), for the 2011/2012 harvest year, two groups were observed for potential evapotranspiration, solar radiation, water balance, and minimum and average temperature, and three groups for precipitation, EVI, and maximum temperature. In 2013/2014, there were two groups for precipitation, solar radiation, water balance, minimum temperature, and three groups for potential evapotranspiration, EVI, and average and maximum temperature. For the 2015/2016 crop year, two groups were identified for precipitation, water balance, and minimum temperature, and three groups were identified for potential evapotranspiration, solar radiation, EVI, average, and maximum temperature.

In the global analysis (Fig. 7), two groups were obtained for the harvest years 2011/2012 and 2013/2014 (Fig. 7(a, b)) and three groups for the harvest year 2015/2016 (Fig. 7(c)). Next, we describe the groupings according to mesoregions of the state of Paraná.

Individual analysis groupings

For the metrics associated with EVI, Group 1, from harvest years 2011/2012 and 2015/2016, presented high values throughout the crop cycle, indicating healthy plant development. However, Group 1, from the 2013/2014 harvest year, exhibited high values only during the sowing stages until grain filling, followed by lower values during the grain-filling stages until soybean maturation, possibly indicating a decline or stress during the final stages of the crop cycle.

Group 2, for all crop years, presented low EVI values during the sowing stages until grain filling (initial stages), and high values during the grain filling stages until soybean maturation (later soybean stages), suggesting slower growth or unfavourable conditions during the early stages, followed by a significant increase in later vegetative development. Group 3, for all crop years, had low EVI values throughout the soybean cycle, indicating possible growth problems or continuous stress over time.

Regarding metrics associated with precipitation, Group 1 demonstrated high values throughout the soybean crop cycle for all crop years, suggesting favourable precipitation conditions over time. Group 2, in turn, presented the opposite, indicating drought conditions or less water availability, potentially negatively affecting the plant growth and yield. Group 3 of the 2011/2012 crop year exhibited low values during the sowing stages until grain filling and high values during the grain-filling period until soybean maturation.

For metrics associated with potential evapotranspiration, Group 1 of the 2011/2012 harvest year presented high values during the initial stages and low values during the later stages, indicating a high initial water demand, followed by a reduction. The opposite pattern was observed in Group 2. Group 1 of the 2013/2014 crop year exhibited low potential evapotranspiration values throughout the cycle, possibly indicating lower humidity or lower water demand. In the 2015/2016 harvest year, Group 1 showed high values of potential evapotranspiration throughout the cycle, whereas Group 2 had low values initially and high values later.

Solar radiation, in turn, presented high values for Group 1 during the initial stages and low values during the later stages of the 2011/2012 and 2013/2014 harvest years. The characterization of this group indicates a greater availability of solar energy during the initial stages of growth, when plants are establishing themselves and actively growing, followed by a reduction in

solar radiation during the stages of grain development and maturation. The opposite occurred for Group 2 during these crop years.

For the 2015/2016 harvest year, Group 1 had high values of metrics associated with solar radiation throughout the crop cycle. With this abundance of solar energy throughout the study period, it is possible to promote healthy plant growth during all phases of the crop. In contrast, Group 2 recorded low values during the sowing stages until grain filling, followed by an increase during grain filling until soybean maturation, indicating a variation in the availability of solar energy throughout the crop cycle. The opposite pattern was observed for Group 3 in this crop year.

Metrics related to water balance showed high values throughout the soybean crop cycle for Group 1 during the harvest years of 2011/2012 and 2013/2014. This characterization suggests abundant availability of water in the soil throughout the period, which can be beneficial for plant growth and development. On the other hand, the opposite was observed for Group 2 during these harvest years as well as for Group 1 in the 2015/2016 harvest year. This situation may indicate drought conditions or reduced water availability in the soil, which can have negative impacts on plant growth and yield.

The minimum temperature metrics revealed distinct patterns between the groups throughout the harvest period. In the 2011/2012 harvest year, Group 1 recorded low values, and Group 2 recorded high values throughout the crop cycle. In the 2013/2014 and 2015/2016 harvest years, an inversion was observed: both Group 1 and Group 2 presented high values during the initial stages until grain filling and low values during filling until soybean maturation, with inversions occurring in Group 2 in 2013/2014 and Group 1 in 2015/2016.

The patterns varied with the average temperatures. In the 2011/2012 and 2015/2016 harvest years, Group 1 recorded low values throughout the cycle, Group 2 presented high values, and Group 3 recorded low values during the 2013/2014 harvest year. However, in the 2013/2014 and 2015/2016 harvest years, Group 1 and Group 2, respectively, exhibited high values during the initial stages until grain filling and low values during filling until soybean maturation, with inversions occurring for Group 2 in 2013/2014 and for Group 3 in 2015/2016.

The maximum temperatures exhibited different behaviours. For the 2011/2012 harvest years, Groups 2 and 3 presented high values throughout the cycle, whereas Group 1 recorded low values. However, in the 2013/2014 and 2015/2016 harvest years, the patterns were reversed: Groups 2 and 3 exhibited low values during the initial stages until grain filling and high values during filling until soybean maturation, with inversions occurring in Group 1 in 2013/2014. These temperature variations at different phenological stages can affect soybean development by subjecting the plants to thermal stress.

Global analysis groupings

Group 1

For all crop years, this group covered most of the North Pioneiro, Central Eastern, Metropolitan Curitiba, Southeast, and Central South mesoregions of the state of Paraná, Brazil (Fig. 7).

The values associated with precipitation and water balance were predominantly high during the 2011/2012 and 2013/2014 harvest years throughout the soybean cycle. The average precipitation ranged from 482 to 827 mm (with a coefficient of variation of 9%) and from 434 to 1062 mm (with a coefficient of variation

of 14%) for the aforementioned crop years, respectively. However, in 2015/2016, precipitation varied between 769 and 1673 mm (with a coefficient of variation of 14%), which is considered relatively low compared to other groups of the same crop year (Supplementary Fig. 1).

Furthermore, negative values were observed in the water balance for 2011/2012 and 2013/2014, indicating a water deficit, and high values for metrics such as EVI, potential evapotranspiration, and solar radiation in 2015/2016, as well as for solar radiation in 2011/2012. The minimum, average, and maximum temperatures remained low throughout the soybean cycle, varying between 10°C and 30°C for the 2011/2012 and 2013/2014 harvest years and between 15°C and 29°C for 2015/2016 (Supplementary Fig. 1).

Group 2

Group 2 for all crop years comprised a large part of the West, Northwest, and Central-West mesoregions (Fig. 7). This group presented low values throughout the soybean crop cycle for metrics associated with precipitation and water balance in crop the 2011/2012 and 2013/2014. Average precipitation ranged from 399 to 720 mm (with a coefficient of variation of 13%) and from 352 to 833 mm (with a coefficient of variation of 16%) for these crop years, respectively, and the water balance showed negative values, indicating a water deficit, especially in the 2011/2012 harvest year. However, for the 2015/2016 crop year, high values were observed for metrics associated with precipitation throughout the soybean cycle, varying between 1073 and 1843 mm (10%) and a positive water balance (Supplementary Fig. 2).

Furthermore, the metrics associated with minimum, average, and maximum temperatures for Group 2 (Fig. 7) remained high throughout the soybean crop cycle, varying between 14°C and 33°C for the 2011/2012 harvest years and 2013/2014, and between 18°C and 32°C for the 2015/2016 harvest year (Supplementary Fig. 2).

On the other hand, during the sowing stages until grain filling, low values were observed for the metrics associated with EVI and solar radiation for the three crop years and for potential evapotranspiration in the 2011/2012 and 2015/2016 crop years, with high values of these metrics during the grain filling stages until soybean maturation.

Cluster 3

For the 2015/2016 harvest year, Group 3 was identified, mainly covering the North Central, Northwest and Central Western mesoregions (Fig. 7(c)).

Throughout the soybean crop cycle, Group 3 presented high values for metrics associated with minimum, average, and maximum temperatures, varying between 17 and 32°C (Supplementary Fig. 3), as well as for metrics associated with EVI.

During the sowing stages until grain filling, high values were observed for metrics associated with precipitation, potential evapotranspiration, solar radiation, and water balance, whereas low values were recorded for these metrics during the grain filling stages until grain maturation.

Differences between groups in the global analysis

One of the main distinctions between Groups 1 and 2 over the three crop years was, is related to the metrics associated with precipitation. Group 1 of the 2011/2012 and 2013/2014 harvest years presented high precipitation values throughout the soybean crop

cycle, whereas the opposite was observed for Group 2 during these periods.

In the 2015/2016 harvest year, for Group 1, low values of metrics associated with precipitation were identified during the sowing stages until grain filling, in contrast with high values during the grain-filling stages until soybean maturation. This trend was reversed in Group 3. Group 2 had high precipitation throughout the soybean crop cycle.

Furthermore, during the reproductive phase from flowering to grain filling, where a minimum of 7 to 8 mm/day of water is required (EMBRAPA SOJA, 2011), Group 1 demonstrated metrics associated with average rainfall below this value, with average of 4.9 mm/day, 5.5 mm/day and 7.0 mm/day, (with coefficients of variation of 19, 13, and 22%) for the harvest years 2011/2012, 2013/2014, and 2015/2016, respectively. Only Group 1 of the 2015/2016 crop year achieved adequate precipitation values. However, there was greater variability in precipitation than in the crop years 2011/2012 and 2013/2014. These variations, together with the lack of adequate precipitation during the critical development phase, could have negative impacts on the productivity of the soybean crop Group 1, during the 2011/2012 and 2013/2014 harvest years.

For Group 2, the metrics associated with average rainfall were even lower, with a daily average of 3.4 mm, 5.7 mm and 18.1 mm (with a coefficient of variation of 30, 19 and 15%) for the harvest years 2011/2012, 2013/2014 and 2015/2016, respectively. These values indicate inadequacy and high variability in the 2011/2012 and 2013/2014 harvest years, respectively. Group 3 of the 2015/2016 crop year recorded an average of 7.7 mm/day, with a variation of 16%, considered adequate for precipitation during these stages. This disparity in precipitation values between harvest years was notable.

Other differences were found in the metrics associated with water balance, which were low for Group 2 in the 2011/2012 and 2013/2014 harvest years, and for Group 3 in the 2015/2016 harvest year. Low water balance values indicate a more severe water deficit in these harvest years, which can negatively affect the growth and development of the crop. In contrast, the other groups presented higher values, suggesting more favourable conditions of water availability throughout the crop cycle.

Regarding the metrics associated with the minimum, average, and maximum temperatures, Group 2 of the three harvest years and Group 3 of the 2015/2016 harvest year presented higher values throughout the entire soybean crop cycle. This is in contrast with Group 1, which had lower temperatures for all three crop years. These variations can influence plant development and soybean crop productivity.

Metrics associated with EVI, potential evapotranspiration, and solar radiation revealed distinct patterns between Groups 2 and 3 in the three crop years. While Group 2 demonstrated low values during the sowing stage until grain filling, followed by high values during the grain filling stage until soybean maturation, the opposite was observed for Group 3 in the 2015/2016 crop year. However, the metrics associated with EVI remained low throughout the soybean cycle. This reversal of patterns in Group 3 of the 2015/2016 harvest year suggests significant differences in the growing conditions between the groups. Increases in potential evapotranspiration, accompanied by an increase in temperature, tend to be harmful to agricultural activities by reducing the water available to plants in the soil (Andrade *et al.*, 2021).

These variations in meteorological variables and EVI have a direct impact on agricultural activities, and consequently, productivity. The groups, as defined by the metrics associated with

meteorological variables and EVI, demonstrated different productivities. Group 1 presented productivity of 2.60 t/ha, 3.09 t/ha and 3.24 t/ha, for the harvest years 2011/2012, 2013/2014 and 2015/2016 respectively (IBGE, 2012, 2014, 2015) (Fig. 7). This tendency to increase productivity throughout the harvest period may be related to the more favourable climatic conditions.

Group 2 productivity was 2.53 t/ha and 2.98 t/ha for the harvest years 2011/2012 and 2013/2014, respectively (IBGE, 2012, 2014), increasing to 3.30 t/ha (IBGE, 2016) in the year –2015/2016 harvest. This increase may be related to more favourable environmental conditions. Group 3 reached approximately 2.58 t/ha in the 2015/2016 harvest year (IBGE, 2016) (Fig. 7).

These data highlight the variability in agricultural productivity over time and the different responses of groups influenced by climate variation and EVI. To obtain a better spatial understanding of productivity in the groups, a Moran scattering map was created (Fig. 7). The map results reveal differences in the distribution of agricultural productivity, especially in the High–High and Low–Low categories.

In the 2011/2012 crop year, around 32.3% of the agricultural areas were classified as High–High and 24.7% as Low–Low (Fig. 7(a)). In 2013/2014, these percentages changed to 26.5 and 21.2%, respectively (Fig. 7(b)). In the 2015/2016 crop year, areas classified as High–High represented approximately 27.2%, while those classified as Low–Low were approximately 24.2% (Fig. 7(c)).

In Group 2, throughout the three harvest years, subregions with low productivity were surrounded by others with equally low productivity. Group 1 had sub-regions with high productivity surrounded by others that were equally productive. This suggests uniformity of high productivity in Group 1 and low productivity in Group 3, characterized by sub-regions with low productivity surrounded by others with equally low productivity (Fig. 7).

These changes in the distribution of areas with different levels of productivity throughout the harvest years may reflect variations in climatic conditions, EVI, and other factors (not evaluated in this study) that affect agricultural productivity. These observations can be useful in understanding how these factors impact agriculture in different regions and periods.

Discussion

Soybean yield is influenced by a complex interaction of factors throughout the crop cycle, with an emphasis on climatic factors. These elements have a direct impact on the growth, development, and production of soybeans, and their interdependence can determine the success or failure of a crop.

Climate interactions can indicate the best periods to plant and harvest soybeans and the most suitable regions for cultivation. The spatial regionalization of Paraná identified areas with favourable climatic conditions for soybeans and those where climatic factors significantly impact productivity. In the three harvests analysed, the mesoregions of Group 1 (Norte Pioneiro, Central Oriental, Metropolitana de Curitiba, Sudeste and Centro-Sul) were more suitable, presenting better productivity (Santos *et al.*, 2024). However, in the 2011/2012 and 2013/2014 harvests, Group 1 had below-ideal precipitation during the reproductive phase (R2 – R5), indicating a possible water deficit, which can reduce productivity due to smaller grain size and weight (Borrmann *et al.*, 2009; Matzenauer *et al.*, 2017). Water deficit increases evaporative demand, leading to faster and more intense stress (Bergamaschi, 2017).

It was also observed that the climatic and agronomic patterns of Groups 1 and 2 in the harvest years 2011/2012 and 2013/2014

were opposite, mainly in relation to precipitation and temperature. These discrepancies can have significant impacts on soybean productivity because the quantity and distribution of water, together with high temperatures, are crucial factors for plant development.

In contrast, Group 2 in the 2011/2012 and 2015/2016 harvest years only showed differences in precipitation and water balance. In 2011/2012, the precipitation and water balance values were below the recommended levels throughout the crop cycle, mainly during the soybean reproductive phase, negatively affecting productivity.

Other factors that may have contributed to the low productivity in Group 2 in 2011/2012 include the occurrence of summers, dry periods with intense heat, strong sunlight, and low relative humidity, which are associated with high temperatures during the soybean maturation phase (Figs 6 and 7), leading to earlier maturation, reduced grain weight and size, and negative effects on productivity (Yee-Shan et al., 2013).

Furthermore, temperatures within the ideal range are conducive to plant growth and productivity. However, even a small increase above threshold levels significantly reduces production yields (Schlenker and Roberts, 2009; Sharma et al., 2022).

In the 2011/2012 harvest year, Paraná faced a lack of rain and high temperatures, especially in the Group 2 mesoregions, owing to the influence of the La Niña phenomenon (CONAB 2012). In contrast, in 2015/2016, there was excess rainfall and low solar radiation due to the El Niño phenomenon, characterizing a wetter year that impacted the agricultural production of soybeans in Paraná. (CONAB 2016)

Variability in soybean productivity is generally accompanied by inter-annual variability in seasonal precipitation. Years with extreme negative precipitation anomalies can result in severe crop loss. This direct association did not occur in the analysis of positive precipitation values. However, excess rainfall during the growing season can have different impacts on yield depending on the stage at which it occurs (Podesta et al., 1999; Llano et al., 2012; Cohen et al., 2021).

It is essential for producers to understand these weather patterns and closely monitor environmental conditions, especially during the soybean reproductive phase, to implement appropriate management strategies and minimize the negative impacts on future harvests. This may include irrigation practices, selection of soybean varieties that are more resistant to environmental stress, and adoption of cultivation techniques that promote plant health and development during this critical period.

The results of this study can still be used to define more appropriate strategies for soybean cultivation in Paraná, providing information on climatic risks for rural producers and responsible bodies, defining regions most suitable for soybean cultivation, and the periods of phenological stages that suffer greater impacts from climate variability. This can guide decision making related to participation in agricultural support programmes, such as rural insurance.

Conclusion

The analysis carried out using multivariate statistics such as EFA and clustering allowed for a deeper understanding of the regional conditions of the state of Paraná in relation to soybean cultivation. Observation of the results at both the individual and global levels revealed similar patterns, although the cartographic representations differed, which enriched our understanding of the variables under study. Regionalization revealed that the mesoregions grouped in Group 1 over the three harvest years showed more favourable climatic conditions for soybean cultivation. On the

other hand, the mesoregions grouped in Group 2 during the 2011/2012 and 2013/2014 harvest years and in Group 3 during the 2015/2016 harvest year faced productivity losses due to unfavourable weather conditions, in addition to other factors that were not addressed in the study. It is essential to consider the impact of climate variability and EVI during the phenological stages of soybean cultivation, particularly during the critical phases of development.

Supplementary material. The supplementary material for this article can be found at <https://doi.org/10.1017/S0021859624000340>.

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References

- Andrade ARS, Silva EG, Vieira AP, Silva MBG, Santos WM and Silva MGS (2021) Factor analysis in the identification of factors for the obtainment of climatological index. *Journal of Environmental Analysis and Progress* **06**, 79–99.
- Anselin L (1995) Local indicators of spatial association -LISA. *Geographical Analysis* **27**, 91–115.
- Baldiseria RS and Dallacort R (2017) Influência das variáveis climáticas declinação solar, fotoperíodo e irradiação no topo da atmosfera em regiões agricultáveis do Brasil. *Revista de Ciências Agroambientais* **15**, 109–115.
- Becker WR, Johann JA, Richetti J and Silva LCA (2017) Data mining techniques for separation of summer crop based on satellite. *Engenharia Agrícola* **37**, 750–759.
- Becker WR, Richetti J, Mercante E, Esquerdo JCDM, Silva Junior CA, Paludo A and Johann JA (2020) Agricultural soybean and corn calendar based on moderate resolution satellite images for southern Brazil. *SEMINA. Ciências agrárias* **41**, 2419–2428.
- Bergamaschi H (2017) ÁGUA. In Bergamaschi H, Bergonci JI (eds). *As Plantas E O Clima: Princípios E Aplicações*. Guaíba: Agrolivros, pp. 256–282.
- Bianchi VR, Oliveira SC, Pinto LB, Andreoli AC and Garnica GB (2016) Grouping of municipalities the middle Paranapanema according to the local agricultural production: a multivariate approach. *Espacios* **37**, 16–26.
- Borrmann D, Junqueira RM, Sinnecker P, Gomes MSO, Castro IA and Marquez UML (2009) Chemical and biochemical characterization of soybean produced under drought stress. *Ciência e Tecnologia de Alimentos* **29**, 676–681.
- Calinski T and Harabasz JA (1974) A dendrite method for cluster analysis. *Communication in Statistics Theory and Methods* **3**, 1–27.
- Charrad M, Ghazzali N, Boiteau V and Niknafs A (2022). Determining the Best Number of Clusters in a Data Set. Repository CRAN.
- Cohen I, Zandalinas SI, Huck C, Fritschi FB and Mittler R (2021) Meta-analysis of drought and heat stress combination impact on crop yield and yield components. *Physiologia Plantarum* **171**, 66–76.
- CONAB – Companhia Nacional de Supply (2012) Monitoring the Brazilian harvest: Grains. 2011/12 harvest. Ninth survey. Agricultural Monitoring. Available at <https://www.conab.gov.br/info-agro/safra> (Accessed July 2020).
- CONAB – Companhia Nacional de Supply (2016) Monitoring the Brazilian harvest: Grains. 2015/16 harvest. Ninth survey. Agricultural Monitoring. Available at (Accessed July 2020).
- CONAB – Companhia Nacional de Supply (2021) Monitoring the Brazilian harvest: Grains. 2020/21 harvest. Ninth survey. Agricultural Monitoring.

- Available at <https://www.conab.gov.br/info-agro/safras> (Accessed June 2021).
- Corea F** (2019) *An Introduction to Data: Everything You Need to Know About AI, Big Data and Data Science*. Berlin: Springer.
- Damásio BF** (2012) Use of exploratory factor analysis in psychology. *Psychological Assessment* **11**, 213–228.
- Davies DL and Bouldin DW** (1979) A cluster separation measure. *IEEE Transactions on Pattern Analysis and Machine Intelligence* **1**, 0–227.
- Doorenbos J and Kassam AH** (1979) *Yield Response to Water*. Rome: Food and Agriculture Organization of the United Nations (FAO irrigation and drainage).
- Dunn J** (1974) Well separated clusters and optimal fuzzy partitions. *Journal Cybernetics* **1**, 95–104.
- ECMWF – European Centre for Medium-Range Weather Forecasts** (2012) Available at http://old.ecmwf.int/about/corporat_brochure/leaflets/Corporate-brochure-2012-en.pdf (Accessed May 2018).
- Eklundh L and Jönsson P** (2015) TIMESAT: a software package for time-series processing and assessment of vegetation dynamics. In Kuenzer C., Dech S and Wagner W (eds.), *Remote Sensing Time Series*. Cham: Springer International Publishing, pp. 141–158.
- EMBRAPA SOJA** (2011) *Soybean Production Technologies – Central Region of Brazil 2012 and 2013*. Brasília: Brazilian Agricultural Research Corporation. National Soy Research Center, 264pp.
- Esquerdo JCDM, Zullo Júnior J and Antunes JFG** (2011) Use of NDVI/AVHRR time series profiles for soybean crop monitoring in Brazil. *International Journal of Remote Sensing. London* **32**, 3711–3727.
- ESRI** (2018) *ArcGIS for Windows Version 10.3*. License type ArcInfo. ESRI - Environmental Systems Research Institute.
- Farias JRB, Nepomuceno AL and Neumaier N** (2007) *Soybean Ecophysiology*. Londrina: EMBRAPA-CNPSo, 9. (EMBRAPA-CNPSo. Technical Circular, 48).
- Fehr WR and Caviness CE** (1977) *Stages of Soybean Development*. Ames: Iowa State University of Science and Technology.
- Ferreira DF** (2018) *Análise Multivariada*, 3rd ed. Lavras: UFLA.
- Fontana DC, Berlato MA, Lauschner MH and Mello RW** (2001) Soybean yield estimation model in the state of Rio Grande do Sul. *Brazilian Agricultural Research* **36**, 399–403.
- Gebert DMP, Kist A and Virgens Filho JS** (2018) Determination of homogeneous rainfall regions in the state of Paraná using multivariate and geostatistical analysis techniques. *Brazilian Journal of Climatology* **23**, 374–388.
- Grzegozewski DM, Johann JA, Uribe-Opazo MA, Mercante E and Coutinho AC** (2016) Mapping soya bean and corn crops in the State of Paraná, Brazil, using EVI images from the MODIS sensor. *International Journal of Remote Sensing* **37**, 1257–1275.
- Grzegozewski DM, Cima EG, Uribe-Opazo MA, Johann JA and Guedes LPC** (2020) Spatial and multivariate analysis of soybean yield in the state of Paraná-Brazil. *Journal of Agricultural Studies* **8**, 387–412.
- Hair JF, Black WC, Babin BJ and Anderson RE** (2014) *Multivariate Data Analysis*, 7th ed. London: Pearson Education.
- Halkidi M, Vazirgiannis M and Batistakis I** (2000) *Quality Scheme Assessment in the Clustering Process. Principles of Data Mining and Knowledge Discovery*. Berlin: Springer-Verlag.
- Hubert LJ and Levin JR** (1976) A general statistical framework for assessing categorical clustering in free recall. *Psychological Bulletin* **83**, 1072–1080.
- IBGE – Brazilian Institute of Geography and Statistics** (2012) Aggregated database - IBGE Automatic Recovery System – SIDRA. Available at: <http://www.sidra.ibge.gov.br> (Accessed Set 2018).
- IBGE – Brazilian Institute of Geography and Statistics** (2014) Aggregated database - IBGE Automatic Recovery System – SIDRA. Available at: <http://www.sidra.ibge.gov.br> (Accessed Set 2018).
- IBGE – Brazilian Institute of Geography and Statistics** (2015) Aggregated database - IBGE Automatic Recovery System – SIDRA. Available at: <http://www.sidra.ibge.gov.br> (Accessed Set 2018).
- IBGE – Brazilian Institute of Geography and Statistics** (2016) Aggregated database - IBGE Automatic Recovery System – SIDRA. Available at: (Accessed Set 2018).
- IBGE – Brazilian Institute of Geography and Statistics** (2018) States Available at: (Accessed September 2018).
- Johann JA, Rocha JV, Oliveira SEM, Rodrigues LHA and Lamparelli RAC** (2013) Data mining techniques for identification of spectrally homogeneous areas using NDVI temporal profiles of soybean crops. *Agricultural Engineering* **33**, 511–524.
- Johann JA, Becker WR, Uribe-Opazo MA and Mercante E** (2016) Use of images from the Modis orbital sensor to estimate dates in the soybean crop development cycle. *Agricultural Engineering* **35**, 1–15.
- Kaiser HF** (1958) The varimax criterion for analytic rotation in factor analysis. *Psychometrika* **23**, 187–200.
- Kaster M and Farias JRB** (2011) Regionalization of VCU tests - cultivation value and use of soybean cultivars. In Embrapa Soy (ed.), *XXXII Soy Research Meeting in the Central Region of Brazil*. São Pedro: Embrapa Soja.
- Krüger CAMB, Fontana DC and Melo RW** (2007) Estimation of soybean grain yield in Rio Grande do Sul using a regionalized agrometeorological-spectral model. *Brazilian Journal of Agrometeorology* **15**, 210–219.
- Krzanowski WJ and Lai YT** (1988) A criterion for determining the number of groups in a data set using sum-of-squares clustering. *Biometrics* **44**, 23–34.
- Li X and Anselin L** (2023) R Library for Spatial Data Analysis. Repository CRAN. <https://CRAN.R-project.org/package=rgeoda>
- Llano MP, Vargas W and Naumann G** (2012) Climate variability in areas of the world with high production of soya beans and corn: its relationship to crop yields. *Meteorological Applications* **19**, 385–396.
- Lopes AR, Marcolin J, Johann JA, Vilas Boas MA and Schuelter AR** (2019) Identification of homogeneous rainfall zones during grain crops in Paraná, Brazil. *Agricultural Engineering* **6**, 707–714.
- MacQueen J** (1967) Some methods for classification and analysis of multivariate observations. In: Proceedings of 5th Berkeley Symposium on Mathematical Statistics and Probability Berkeley: University of California Press.
- MAPA – Ministry of Agriculture, Livestock and Supply** (2022). Good agricultural practices for the production of safe food. Secretariat of Innovation, Sustainable Development and Irrigation. Available at <https://www.gov.br/agricultura/pt-br/assuntos/sustentabilidade/boas-praticas-agricolas/publicacoes-tecnicas/livro-boas-pratica-agricolas-para-a-producao-de-alimentos-seguros.pdf> (Accessed May 2024).
- Mathur M** (2015) Spatial autocorrelation analysis in plant population: an overview. *Journal of Applied and Natural Science* **7**, 501–513.
- Matzenauer R, Radin B and Maluf JRT** (2017) O fenômeno ENOS e o regime de chuvas no Rio Grande do Sul. *Agrometeoros* **25**, 323–331.
- NASA – National Aeronautics and Space Administration** (2019). Technical specifications: Moderate resolution imaging spectroradiometer (MODIS). (Accessed September 2019).
- Nascimento L JR, Silvestre MR and Neto JLS** (2020) Trends and rainfall tropicalization in Paraná State, south of Brazil. *Atmosfera* **33**, 1–18.
- Neumaier N, Nepomuceno AL and Farias, JRB** (2021) *Climate Risk Zoning for Soybeans*. Brasília, DF: Embrapa Soja. Available at <https://www.embrapa.br/agencia-de-informacao-tecnologica/cultivos/soja/pre-producao/caracteristicas-da-especie-e-relacoes-com-o-ambiente/zonamento-agroclimatico/zoneamento-de-risco-climatico-a-soja> (Accessed May 2024).
- Oliveira JRT and Padovani CR** (2017) Analysis of the interrelationship of agricultural productivity and climatic characteristics in the southeast region of the State of Mato Grosso, using multivariate techniques. *Engineering and Science* **2**, 1–12.
- Paludo A, Becker WR, Richetti J, Silva LCA and Johann JA** (2020) Mapping summer soybean and corn with remote sensing on Google Earth Engine cloud computing in Parana state – Brazil. *International Journal of Digital Earth* **13**, 1624–1636.
- Pereira AR, Angelocci LR and Sentelhas PC** (2002) Agrometeorology: fundamentals and practical applications. *Agricultural Journal* **1**, 478.
- Podesta GP, Messina C, Grondon M and Magrin G** (1999) Associations between grain crop yields in central-eastern Argentina and El Niño-Southern Oscillation. *Journal of Applied Meteorology* **38**, 1488–1498.
- Raiche G, Walls TA, Magis D, Riopel M and Blais JG** (2013) Nongraphical solutions for Cattell's scree test. *Methodology* **9**, 23–29.
- R Development Core Team** (2022) *R: A Language and Environment for Statistical Computing*. Version 4.0.2. Vienna: R Foundation for Statistical Computing.

- Ribeiro RC, Dallacort R, Barbieri JD, Santi A and Ramos HC** (2015) Zoning of the annual water balance of sugar cane for the state of Mato Grosso. *Biosphere Encyclopedia* **11**, 1–13.
- Richetti J, Cima EG, Johann JA and Uribe-Opazo MA** (2018) Data mining techniques for rainfall regionalization in Parana state. *Acta Iguazu* **7**, 1–8.
- Rousseeuw P** (1987) Silhouettes: a graphical aid to the interpretation and validation of cluster analysis. *Journal of Computational and Applied Mathematics* **20**, 53–65.
- Santos NC, Correia AB, Behling SA and Feiden A** (2024) Análise da produção, produtividade e valor das culturas da soja, milho e trigo por mesorregião no Paraná. *Revista Desenvolvimento em Questão* **60**, 1–20.
- Sarle WS** (1983) *SAS Technical Report A-108, Cubic Clustering Criterion*. Cary: SAS Institute Inc.
- Sathiaraj D, Huang X and Chen J** (2018) Predicting climate types for the Continental United States using unsupervised clustering techniques. *Environmetrics* **25**, 1–12.
- Savitzky A and Golay MJE** (1964) Smoothing and differentiation of data by simplified least squares procedures. *Analytical Chemistry* **36**, 1627–1639.
- Schlenker W and Roberts MJ** (2009) Non-linear temperatures effect indicate severe damage to US crops un-der climate change. *Proceedings of the National Academy of Sciences of the USA* **106**, 37.
- Scott AJ and Symons MJ** (1971) Clustering methods based on likelihood ratio criteria. *Biometrics* **27**, 387–397.
- Serio L, Spescha L and Murphy GM** (2006) Validación de las precipitaciones decádicas de la región pampeana na estimadas por el modelo atmosférico del ECMWF. p.45–52. In: *2° International Workshop on Crop Monitoring and Forecasting in South America, 2006. Proceedings...* Montevideo: South America Scientific Network on Crop Monitoring and Forecasting.
- Sharma RK, Kumar S, Vatta K, Bheemanahalli R, Dhillon J and Reddy KN** (2022) Impact of recent climate change on corn, rice, and wheat in south-eastern USA. *Scientific Reports* **12**, 16928.
- Shirin AHS and Thomas R** (2016) Regionalization of rainfall in Kerala state. *Procedia Technology* **24**, 15–22.
- Souza CHW, Mercante E, Johann JA, Lamparelli RAC and Uribe-Opazo MA** (2015) Mapping and discrimination of soya bean and corn crops using spectro-temporal profiles of vegetation indices. *International Journal of Remote Sensing* **36**, 1809–1824.
- Stansel JW and Fries RE** (1980) Agrometeorological variables in rice physiology relationships Moisture Water is the main determinant of rice cropping and yield. Moisture amounts: In *Proceedings of a Symposium on the Agrometeorology of the Rice Crop* (p. 201).
- Taiz L and Zeiger E** (2009) *Plant physiology*, 4th ed. Porto Alegre: Artmed, pp. 848.
- Thomas AL** (2018) *Soy. Types of Plant Growth. Sector Library of the Faculty of Agronomy*. Porto Alegre: UFRGS.
- Thornthwaite CW and Mather JR** (1955) The water balance. Centerton: Drexel Institute of Technology. *Climatology* **8**, 1.
- United States Department of Agriculture – USDA** (2021) Soja. Available at (Accessed Jul 2021).
- Verica WR** (2018) Semi-automatic mapping using a spectro-temporal pattern of agricultural areas and permanent targets with EVI/MODIS in Paraná (Masters Dissertation). Paraná: State University of Western Paraná.
- Yee-Shan K, Wan-Kin AY, Yuk-Lin Y, Man-Wah L, Chao-Qing W, Xueyi L and Hon-Ming L** (2013) Drought stress and tolerance in soybean. In Board J (ed.), *Comprehensive Survey of International Soybean Research: Genetics, Physiology, Agronomy and Nitrogen Relationships*. Louisiana State University, London, UK: IntechOpen Publisher. <http://dx.doi.org/10.5772/52945>