

# Evaluation of time-series of satellite reflectance data for land delimitation using clustering algorithms



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## Motivation

- Identification of **homogeneous zones of crops based on remoted sensed reflectivity**
- Study how the temporal resolution affects to the delimitation of zones
- Apply it to Terras Gauda grapes crop.



Terras Gauda vineyard

## PAM Clustering Algorithm

- It is a **partitioning algorithm**. Thus, it breaks the input data up into groups until some stability condition is reached.
- The number of groups is defined in advance.
- PAM stands for **Partition Around Medoids**. It tries to find a set of objects called medoids that are centrally located in clusters.
- PAM is an algorithm more robust than K-means because it minimizes a sum of dissimilarities instead of a sum of squared euclidean distances.
- Clustering performance metric: **Silhouette coefficient**,  $s(i) = \frac{b(i) - a(i)}{\max\{b(i), a(i)\}}$ , with  $a(i)$  = average dissimilarity between  $i$  and all other points of the cluster to which  $i$  belongs and  $b(i) = \min_{C \neq D} \{d(i, C)\}$ ,  $\forall$  cluster  $C$
- Dissimilarity metric: **Manhattan distance**

## Digital Data

- Collected by **MODIS** (Moderate Resolution Imaging Spectroradiometer) .
- Negative values correspond to water, clouds or snow since their reflectance in the visible spectrum is greater than the corresponding in near infrared, whilst soil and rocks have values near zero.
- Ranges between 0.1 and 0.6 are indicators of vegetation.
- Values above 0.6 correspond to dense vegetation canopy.

| Name    | Data Product                            | Res. (m) | Frequency |
|---------|---|----------|-----------|
| MYD09GA | Surface Reflectance Bands 1-7           | 500 m    | Daily     |
| MOD09GQ | Surface Reflectance Bands 1-2           | 250 m    | Daily     |
| MOD11A1 | Land Surface Temperature and Emissivity | 1000 m   | Daily     |
| MOD13Q1 | Vegetation Indices                      | 250 m    | 16 days   |
| MOD15A2 | Leaf Area Index - FPAR                  | 1000 m   | 8 days    |
| MOD14A1 | Thermal Anomalies and Fire              | 1000 m   | Daily     |
| MOD44B  | Vegetation Continuous Fields            | 250 m    | Annual    |

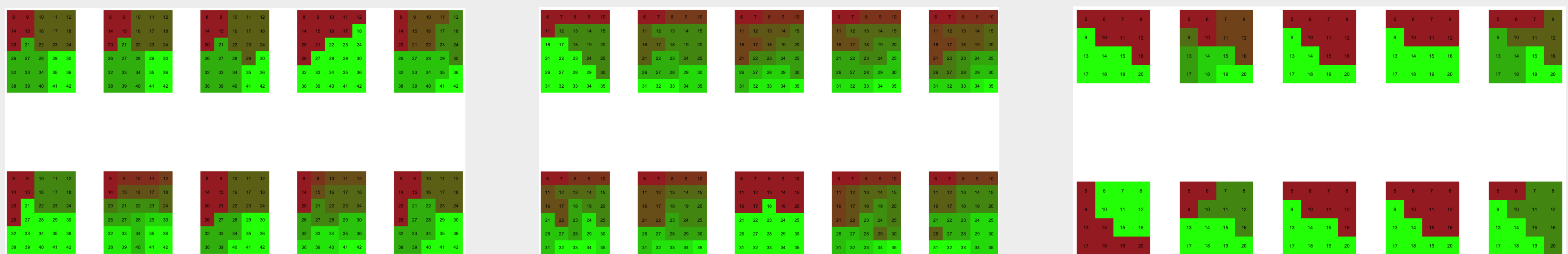
Table 1: A sample of MODIS data products related to Precision Agriculture

| Column           | Description  |
|------------------|--|
| x                | Coordinate x of the data point in the UTM 29 CRS               |
| y                | Coordinate y of the data point in the UTM 29 CRS               |
| date             | Year + Day number of the year in the format YYYYddd            |
| refl_b01         | Reflectivity values from MOD09GQ band 1                        |
| refl_b02         | Reflectivity values from MOD09GQ band 2                        |
| num_observations | The number of observations for this measure                    |
| QC_250m          | A byte of information about the quality of the measure         |
| NDVI             | Normalized Difference Vegetation Index                         |
| NDVI_scaled      | Normalized Difference Vegetation Index. Scaled [0..255] values |

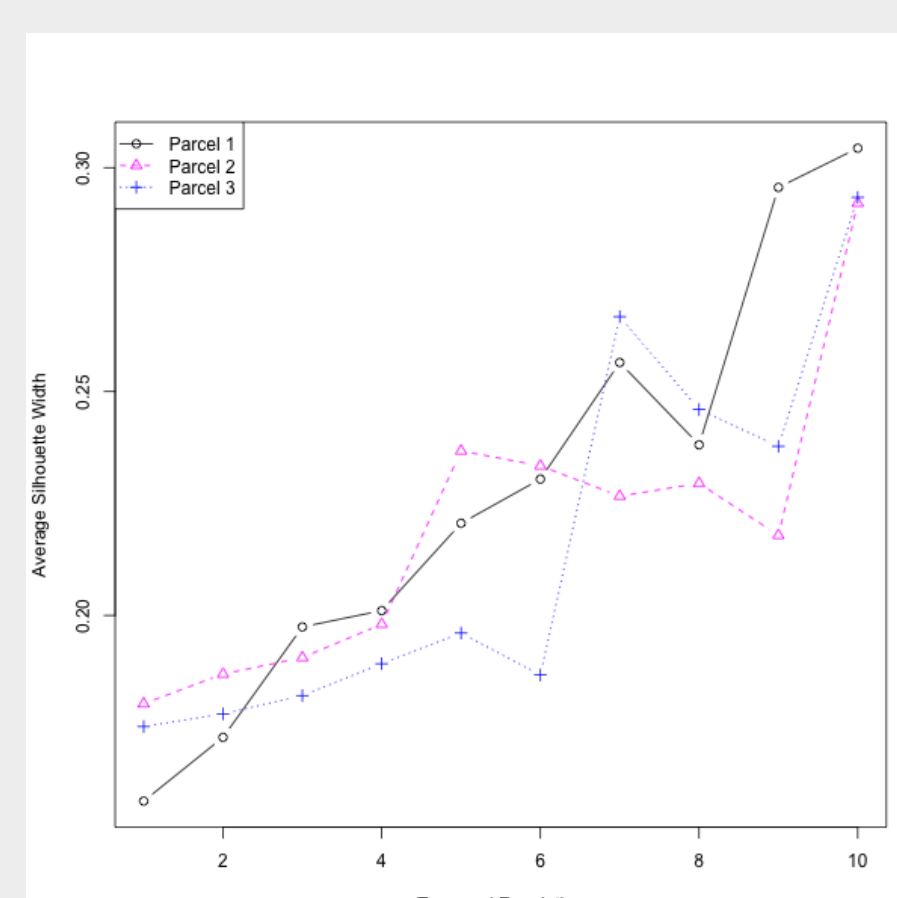
Table 2: Columns of the dataset for the MOD09GQ data product. Spatial resolution: 250 m. Temporal resolution: daily. CRS: UTM 29

## Experiments

- **Terras Gauda** vineyard is divided into three separated parcels ( $p_1, p_2, p_3$ ).
- $p_1$  and  $p_2$  are represented using 30 points and  $p_3$  by 16 points (according to MODIS resolution).
- Each point is characterised by **4 variables per day** (surface reflectance for spectral bands 1 and 2, NVDI and NVDI-scaled indexes).
- As **data** were extracted for **90 days**, each point  $x$  is represented by **360 attribute** as follows  $x = (b1_{day_1}, b2_{day_1}, NVDI_{day_1}, b1_{day_2}, b2_{day_2}, NVDI_{day_2}, \dots, b1_{day_{90}}, b2_{day_{90}}, NVDI_{day_{90}})$
- Points are clustered using PAM algorithm using manhattan distance as similarity metric.
- The optimum number of clusters was computed according to the Silhouette coefficient.



From left to right figures respectively show the land delimitation for parcels  $p_1, p_2$  and  $p_3$ . The structure of each figure is the following: It contains 10 squares divided into smaller squares according to the the spatial resolution provided by MODIS. Considering each figure as a matrix with 2 rows and 5 columns, the square at position  $[1, j]$  contains the clusters obtained when temporal resolution is  $j$ . The square at position  $[2, j]$  contains the clusters obtained when temporal resolution is  $j + 5$ . Therefore the topleft square represents clustering results for daily resolution and the bottomright square represents clustering results for a ten days resolution. Each cluster is represented by a different color.



Silhouette coefficient. X-axis represents the temporal resolution  
Y-axis represents the value of the coefficient