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Pattern Mining in Sentinel 2A Satellite Images Using Knime Analytics Platform

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Introduction

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> The importance of research is to allow image processing without human interference to efficiently determine land use and land cover changes.

> Land use patterns and land cover mining tool designed to manage knowledge from Sentinel 2A series satellite image data.

> It was elaborated on Knime Analytics platform through the selection of configured and connected nodes constituting a workflow composing all the methodological phases.

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Introduction

> The graphic platform allows different configurations for the nodes with the main phases: loading and visualization of images; feature extraction (non-redundant numeric vector) that characterizes land use and cover; creation of attributes for each target (class of land use and land cover; test phase and evaluation of the predictive model.





Introduction

> The peculiarity in this case is that we must extract a vector of numerical resources from the image before starting the machine learning of the classifier algorithm in the implementation phase.

> Two reasons justify:

volume of images available
 is a variety of tools that are easy to use for data mining



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Material and Methods

- 1. *Images dataset:* 1,370 files with 30 files per spectral band with 10 X 10 pixels of 10 spectral bands
- 2. *Building the Workflow in Knime:* was elaborated by placing nodes with their different functions and interconnected with each other to flow processing in main steps
 - . preparation of data such as reading, extracting characteristics and filtering data;
 - . data partitioning, machine learning in the decision tree and random forest algorithms and classification prediction;
 - . performance analysis of the classification algorithms.





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Data sets:

1.370 images Tiff format file

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Building the workflow in Knime Analytics Platform, partial view

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Material and Methods

3. *Feature extraction:* was performed by the node called "Image Features" and consisted of configuring extraction options

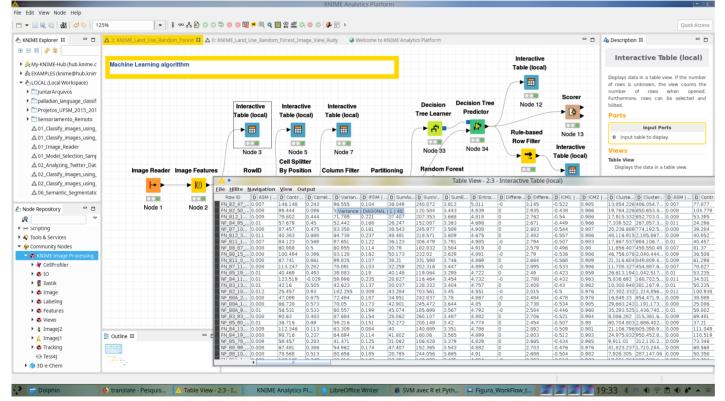
4. Creating the target attibute and learning and evaluating th predictive model





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Feature extraction of images:

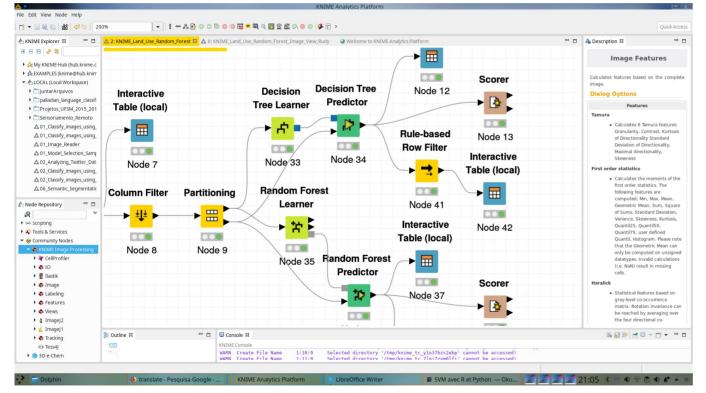
. values minimum, maximum, average, geometric mean, statndard deviation, variance, contrast, correlation, entropy and other parameters



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Data set was partitioned into two other sets, training and testing

The data partitioning had two flows:

1. Decision Tree Predictor algorithm;

2. Randon Forest Predictor algorithm



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Results and Discussion

1. *Decision Tree algorithm:* the node produces two information to analyze the classifier's performance.

The first consists of a graphical information in the form of a tree in which it presents the weights and percentages of each of the parameters, used in the characterization process of data image.

The second possible information is represented which shows the performance values of the classifier such as: samples classified correctly, classification errors and accuracy.

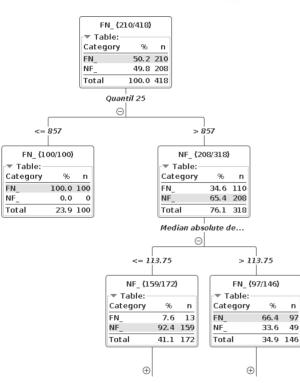
The Decision Tree classification algorithm had an accuracy of 87.778%, an error (incorrectly classified samples) of 12.22%, with the Cohen's Kappa k = 0.756 index, considered satisfactory. Federal University of Santa Maria





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The **left figure** represent the classifier performance with weights and percentages of each of the parameters.

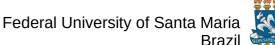
The right figure present Confusion Matrix.





Results and Discussion

2. Random Forest algorithm: shows the performance values of the classifier such as: samples classified correctly, classification errors and accuracy. The Random Forest classification algorithm had an accuracy of 93.333% accuracy, an error (incorrectly classified samples) of 6.667%, with the Cohen's Kappa k = 0.867 index, considered satisfactory.





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CHALLENGES IN A NEW REALITY

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Overview of Workflow Knime for image classification

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Conclusion

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1. The Knime Analytics platform appears as a high-performance tool for complex analyzes without requiring a single line of code with a programming language;

2. The flow model used allows it to be improved, as it is possible to export, edit and adapt it to the interests of each user.

3. The analysis of the performance of the Decision Tree and Random Forest algorithms allowed us to conclude that it is possible to classify the images with the necessary precision.

4. Random Forest was the one that presented the best performance in the classification of images with the target of interest Forests.

5. Then, it is concluded that it is possible to make an intelligent knowledge management.



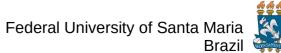
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Thanks for your attention!

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