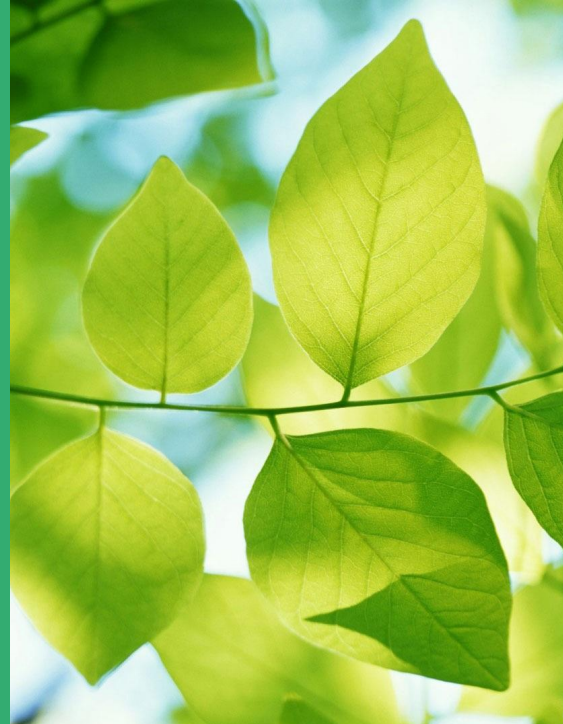


LEAVES CLASSIFICATION

PROJECT EDUPARK

COMPUTER VISION



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Teachers António Neves and Paulo Dias

Group 9

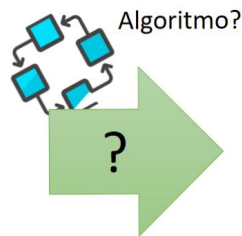
11 January 2017

CONTENTS

1. Main Goals
2. Technologies
3. Segmentation and processing
4. Extracted Features
5. Classifier
6. Training Data
7. Testing Data
8. Results
9. Demo
10. Conclusions
11. Future work

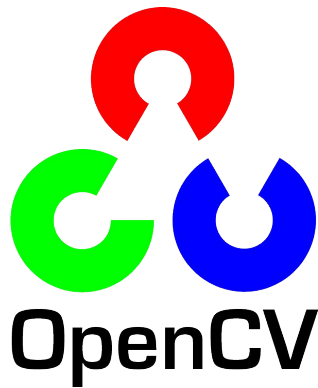
MAIN GOALS

- Receiving a picture of a leaf predict its species name
- Test multiple features to compare performance and have a perception of what the best features to work with
- Train our own dataset with a classifier



Base de dados com
características

TECHNOLOGIES

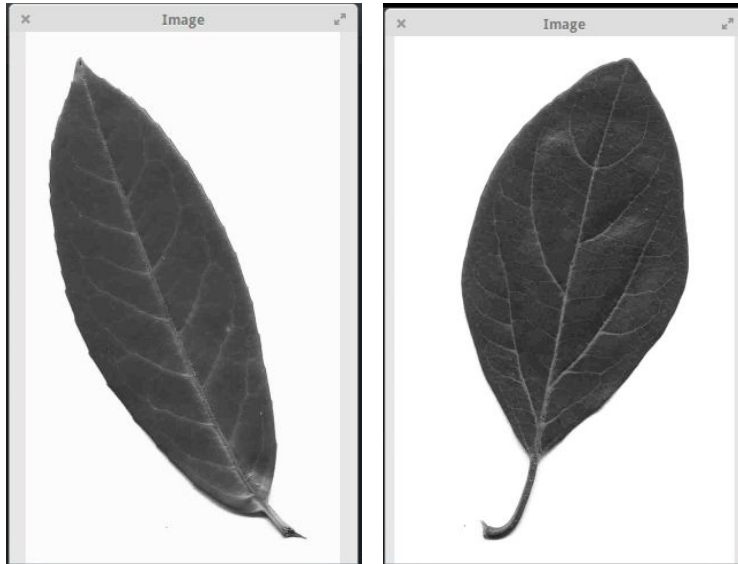


SEGMENTATION AND PROCESSING (I)



- Necessary procedure to **minimize noise**
- **Objective:** to obtain leaf contour
- **Example input:** two leaves

SEGMENTATION AND PROCESSING (II)



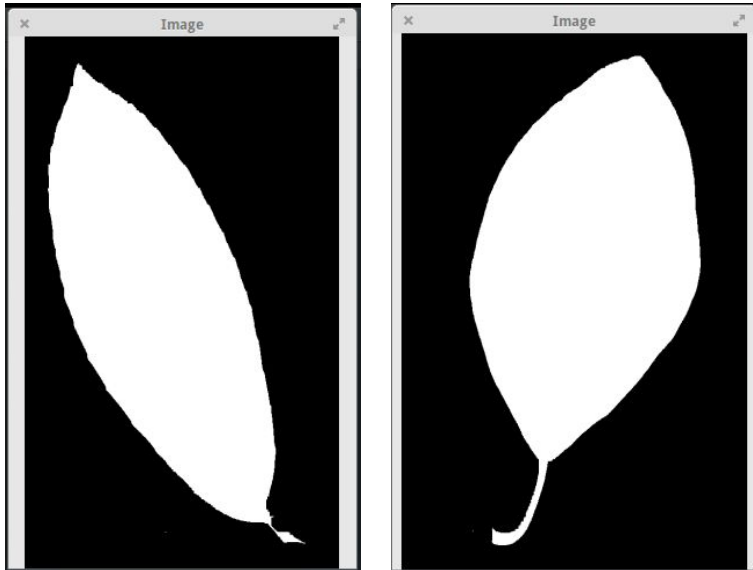
- Necessary procedure to **minimize noise**
- **Objective:** to obtain leaf contour
- **Example input:** two leaves
- Convert to **grayscale**

SEGMENTATION AND PROCESSING (III)



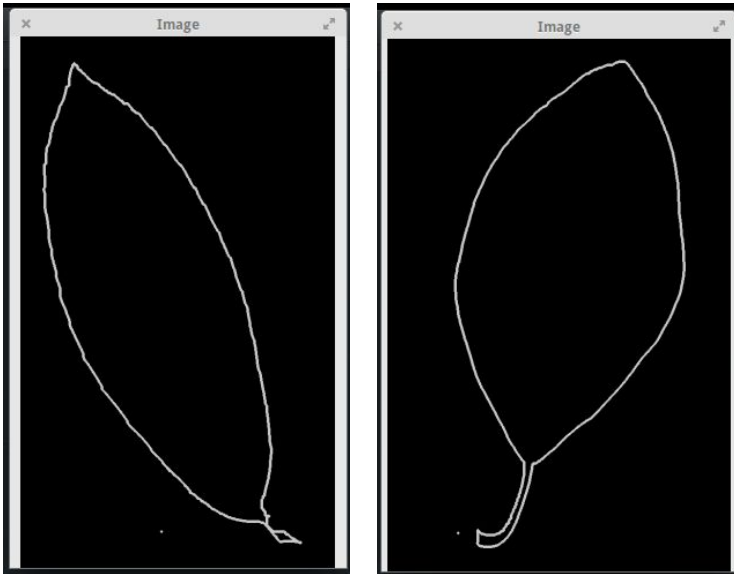
- Apply **threshold**
- **Inverted binary**
- **With** some noise...

SEGMENTATION AND PROCESSING (VI)



- Apply **morphological transformations**
- **Closing**
- Structuring element: **kernel**
- Allows to **remove noise**

SEGMENTATION AND PROCESSING (V)



- Find and draw contours
- `findContours()`
- `drawContours()`

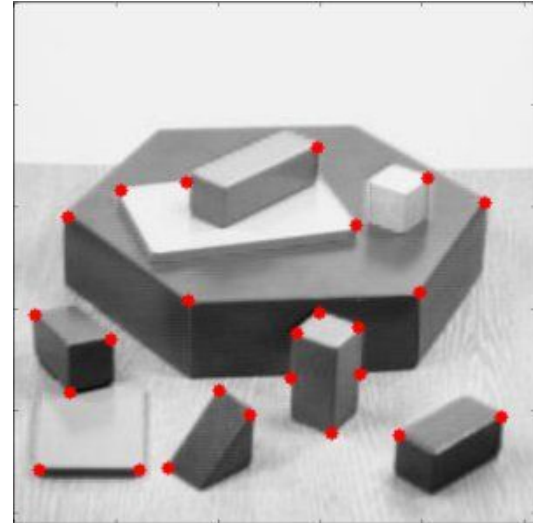
SEGMENTATION AND PROCESSING (VI)



EXTRACTED FEATURE

NUMBER OF CORNERS

- `cv2.goodFeaturesToTrack()`
- Used variation of the **Harris Corner Detection** algorithm
- Determines **strong corners** on an image.

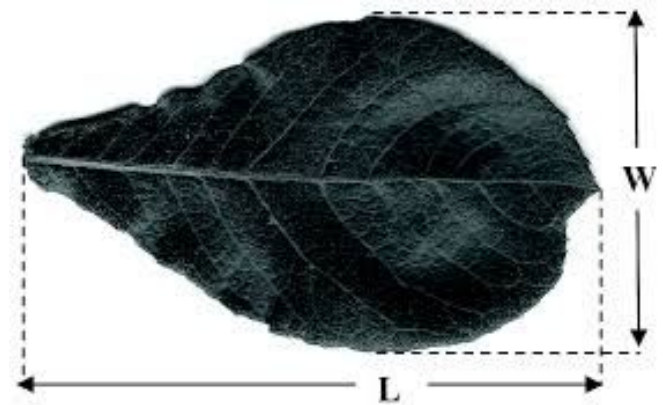


EXTRACTED FEATURE

RATIO OF LENGTH AND WIDTH

- Only width and height are **not useful!**

$$\textit{AspectRatio} = \frac{\textit{Width}}{\textit{Height}}$$



EXTRACTED FEATURE

RATIO OF PERIMETER AND AREA

- Only areas and perimeters are not useful!

$$rpa = \frac{\textit{perimetro}}{\textit{area}}$$

rpa: razão entre perímetro e área



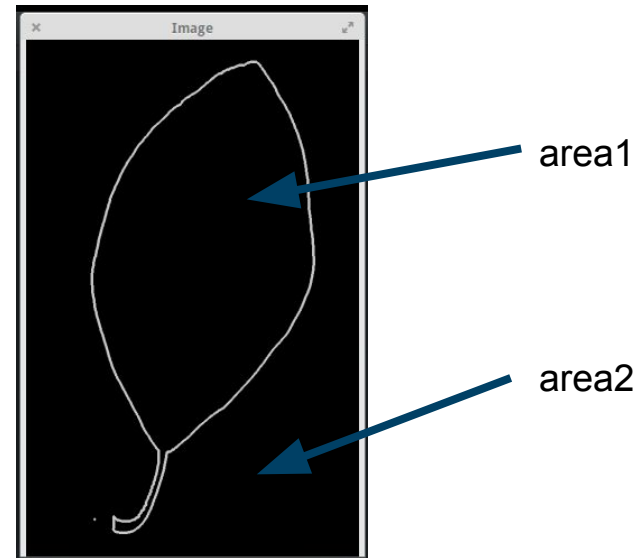
EXTRACTED FEATURE

RATIO OF AREA

- Ratio between **area of leaf** and the area of **outside the leaf's** contourn;

$$ra = \frac{area1}{area2}$$

ra: razão entre áreas

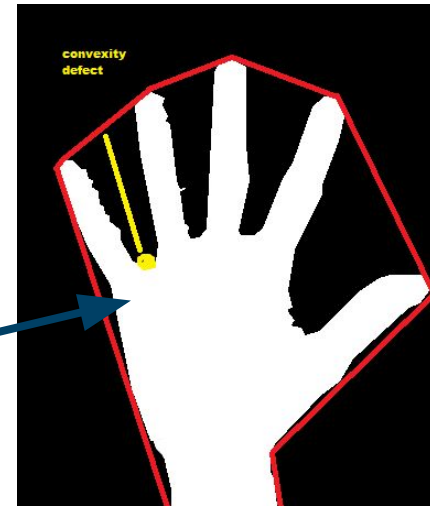


EXTRACTED FEATURE

SOLIDITY

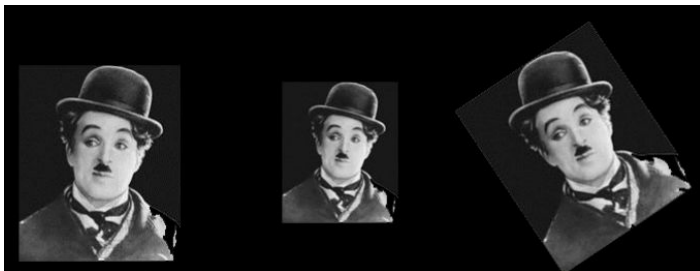
- Ratio between the **contour area** and the **convex area** formed by the union of the points forming a polygon

$$Solidity = \frac{ContourArea}{convexHullArea}$$



EXTRACTED FEATURE MOMENTS

- Help us to calculate some **characteristics** of the image such as, **mass center** or **orientation** of the image.
- `cv2.moments(image)`



```
ana@chicky: ~/Dropbox/Code
File Edit View Terminal Help
* Contour[26] - Area (M_00) = 263.00 - Area OpenCV: 263.00 - Length: 174.97
* Contour[27] - Area (M_00) = 252.00 - Area OpenCV: 252.00 - Length: 169.80
* Contour[28] - Area (M_00) = 242.00 - Area OpenCV: 242.00 - Length: 62.63
* Contour[29] - Area (M_00) = 230.00 - Area OpenCV: 230.00 - Length: 60.28
* Contour[30] - Area (M_00) = 382.00 - Area OpenCV: 382.00 - Length: 73.94
* Contour[31] - Area (M_00) = 364.00 - Area OpenCV: 364.00 - Length: 71.60
* Contour[32] - Area (M_00) = 392.50 - Area OpenCV: 392.50 - Length: 75.36
* Contour[33] - Area (M_00) = 374.50 - Area OpenCV: 374.50 - Length: 73.01
```

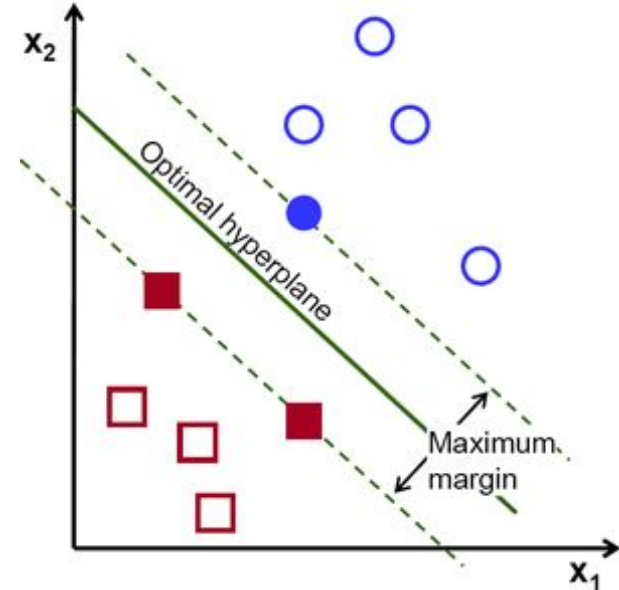

EXTRACTED FEATURE SUMMARY

- Number of Corners
- Ratio of length and height
- Ratio of Perimeter and area
- Ratio of area
- Solidity
- Moments

CLASSIFIER

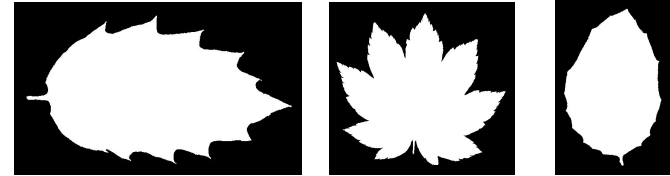
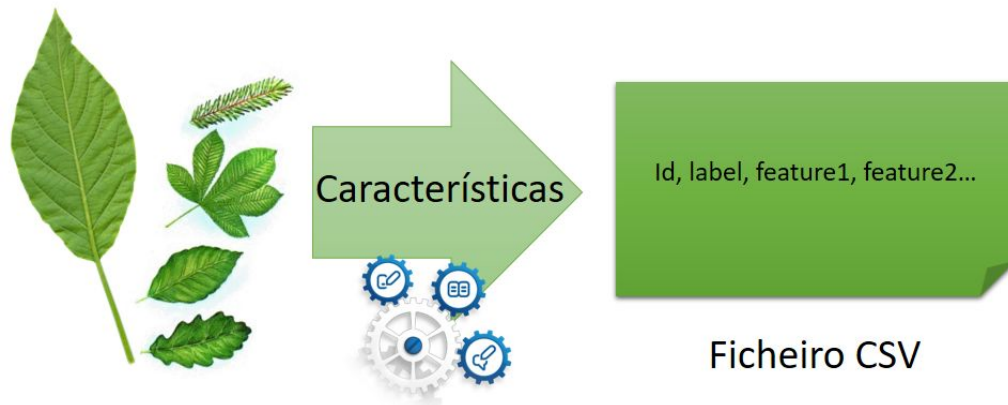
SUPPORT VECTOR MACHINE (SVM)

- Supervised learning models with associated learning algorithms that analyze data used for classification and regression analysis.



TRAINING DATA (I)

- **≈ 160** leaf images (binary images)
- **10** species
- Output: **CSV file** with the extracted features



TRAINING DATA (II)

- `python training_Leaves.py` or `python training_Leaves.py nm`
Extracts all the characteristics **except the moments**
- `python training_Leaves.py all`
Extracts **all the features**

	id	label	number_corners	length_width_ratio	perimeter_area_ratio	ratio_of_areas	solidity	mu02	mu03
2	1	Acer_platanoides	22	1.60103626943	0.403494207766	55.1340138108	0.523424099838	53324969550.7	6.64652797279e+12
3	2	Acer_platanoides	53	1.63001240182	0.411419491143	86.1167319742	0.447832445344	611007314359.0	3.1631336848e+14
4	3	Acer_platanoides	27	1.28092783505	0.409558004209	54.570756469	0.522850288505	47910792572.0	1.00351232489e+12
5	4	Acer_platanoides	29	1.61773255814	0.402451920492	54.8989299733	0.496183423517	45808535982.0	6.13111977024e+12
6	5	Acer_platanoides	33	1.63769690386	0.405251007552	72.3952341824	0.556717157664	208447567256.0	2.56630831266e+13

TESTING DATA (I)

- `python test_Leaves.py`
All the features **except the moments** are used in
classification
- `python test_Leaves.py all`
All the characteristics are used in the **classification**

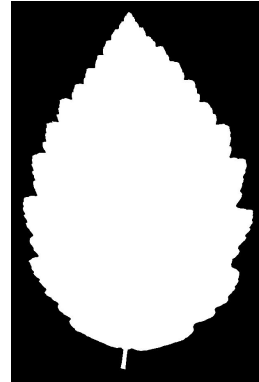


TESTING DATA (II)

- Test image in folder **Test_Leaves**
- Extract the characteristics of testing images
- Send those and training data's characteristics to the SVM
- Predict the plant's specie

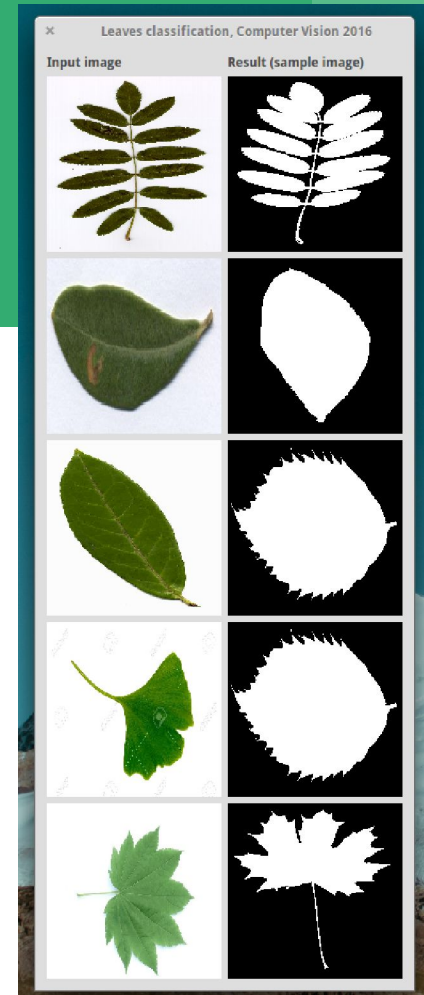


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RESULTS

- Interface with
 - input images (at left)
 - predicted specie image (at right)
- In terminal we have the name of the predicted specie and of the input image (e.g Ginkgo_biloba2.jpg)



DEMO

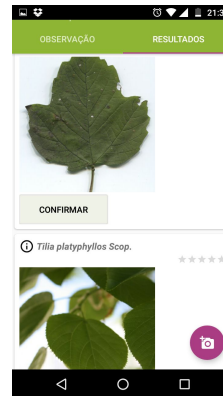
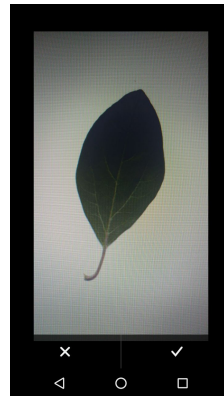
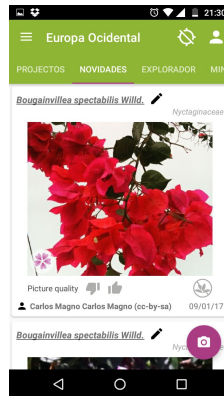
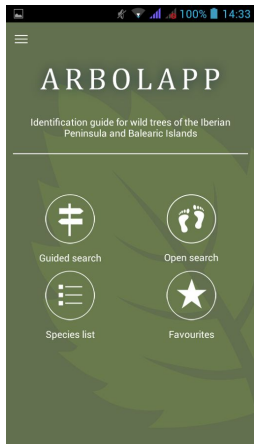
CONCLUSIONS

- Previsions are **not** always correct.
- SVM predict with **higher accuracy** testing the features **without moments**.
- Input images **used in training** are always predicted **correctly**.

FUTURE WORK

- Use **deep learning** and/or **neural networks** to increase accuracy.
- Increment/Improve the **vector of features**.
- **Increase** the training **data set**, maybe with colored images too.
- **Improve** the predictions for images in a more **realistic context**, maybe with leaf tracking first and then all the pré-processing

MOBILE APPLICATION FOR LEAVES RECOGNITION



QUESTIONS?

