Plant leaf classification

Project Milestone

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1 Introduction

We have considered the following approaches after literature study.

- Local descriptors + Bag of Features (BoF)
 Due to the simplicity and performance, this well established approach was taken at first. Interest points are detected from the raw images and then local invariant feature descriptors are collected, which are clustered to form the visual vocabulary/codebook. Afterwards, each raw image can be represented with histograms of visual words, i.e. term vectors.
- 2. Global shape feature descriptors Instead of describing the patches of image, the image, as a whole, can be represented with a global description. In the context of plant classification, the shape is considered the most distinct characteristics [3], among other types of descriptions, i.e., color and texture.

In both cases, the new image representations can be used to train classifiers of our choices.

2 Data set

We found these datasets.

- 1. UCI leaf dataset [5]: 40 species with 5 to 16 samples per specie
- 2. Kaggle leaf dataset[6]: 99 species with 16 samples per specie
- 3. Swedish leaf dataset [7]: 15 species with 952 samples (roughly 60 samples per specie)
- 4. Flavia leaf dataset [8]: 33 species with roughly 60 samples per specie
- 5. Pl@ntNet [9]: the most challenging dataset as images are collected through crowd sourced application. 71 species with 5436 images.

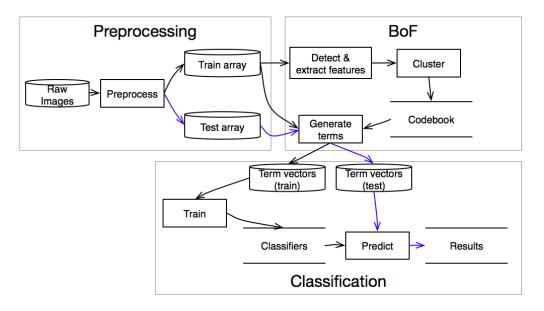


Figure 3.1: System Design for BoF

Table 3.1: preliminary test accuracy of BoF approach on Swedish dataset

Vocabulary size	Softmax	Linear SVM	RBF SVM
100	0.71	0.86	0.87
1000	0.79	0.93	0.93

3 Status

We have prototyped a BoF system, based on OpenCV package, and test with the Swedish dataset [7]. Here is the illustration of the system 3.1. During preprocessing, raw images are converted to gray-scaled images and resized to reduce computation complexity. We extract SIFT descriptors from the pixels after detecting the key points. Limiting the width of image to 128 pixels, we have roughly 200 SIFT descriptors per image. Then all the descriptors are clustered to build visual words via K-Means. Due to computation complexity, we pick randomly 100 training images to build the visual vocabulary for the initial run. Here are the preliminary results 3.1

We noted that

- 1. Since there are relative large features per training sample, we are not surprised that SVM with linear kernel performs well. However, there seems to be overfitting for Softmax.
- 2. The test accuracy is improved with larger vocabulary size. This makes sense since more local feature descriptors could represent the discriminative characteristics better

and hence better accuracy.

4 TODO

- 1. We will implement the global descriptor approach. We then compare the pro and con between these two approaches.
- 2. Swedish data set is high quality dataset with large samples per species and we will test our BoF system with more challenging data set. We expect worse test results and this will give us motivation to enhance our system.
- 3. We reduce the total number of local descriptors for build the vocabulary by limiting the samples. We plan to use the dimension reduction techniques (e.g. PCA) for such task.

References

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