

# SVM, Decision Trees and Ensemble methods

## 1 SVM with scikit-learn

We use the object `sklearn.svm.SVC` :

```
from sklearn.svm import SVC
```

1. Use the website :

<http://scikit-learn.org/stable/modules/svm.html>

and the dataset Iris. Implement a classifier which classifies class 1 against class 2 of the dataset `iris` using the two first variables and a linear kernel. Use half of the dataset for training and half of the dataset for validation. To import the `iris` dataset, type

```
from sklearn import datasets
iris = datasets.load_iris()
X = iris.data
y = iris.target
X = X[y != 0, :2]
y = y[y != 0]
```

2. Split into train set and test set
3. Compare the result of a linear SVM with a SVM based on polynomial kernel. One can use the function `score` to compare the different models
4. Evaluate the impact of the choice of the kernel and the regularisation parameter  $C$ .
5. Improve the robustness of the results with cross validation

[https://scikit-learn.org/stable/modules/cross\\_validation.html](https://scikit-learn.org/stable/modules/cross_validation.html)

## 2 Decision trees and ensemble methods

This section aims at using Ensemble Methods on real and synthetic data using the package `scikit-learn`.

### 2.1 Data importation and preparation

1. Import the dataset `digits` from the library `datasets`
2. Define the data and the labels
3. Split the dataset into train set and test set

## 2.2 Decision Trees

1. Import `scikitlearn` and the library `tree` of `sklearn` appropriate for decision tree
2. Fit a decision tree on the train set
3. Performance of this model on the test set?

## 2.3 Bagging, Random Forest and Boosting

1. Import `BaggingClassifier`, `RandomForestClassifier` and `AdaBoostClassifier` from the library `sklearn.ensemble`
2. Compare the performance of these methods on the dataset `digits`