SPLEX TME 4

Decision Trees and Random Forest

The goal of the TME is to develop practical skills to use decision trees and random forest for real biological applications.

We will use the *scikit-learn Python* library `http://scikit-learn.org` which is already installed on the computers.

Data

- Diabetes Remission Prediction. The problem is to predict whether a diabetic patient will resolve or will not resolve his diabetes after a gastric bypass surgery.
  1. `patients_data.txt` – Observations: 200 patients, 4 clinical variables: age of patients (continuous), HbA1C (continuous), insuline taken (categorical, yes or not), other anti-diabetic drugs are taken (categorical, yes or not)
  2. `patients_classes.txt` – Classes: 0 (Diabetes Remission) and 1 (Non-Remission) for 200 patients

Libraries

You will need to load the following packages:

```
import pandas as pd
import graphviz
from sklearn import tree
from sklearn.ensemble import RandomForestClassifier
```

Analysis

Read the data

```
data_diabetes = pd.read_table('patients_data.txt',sep='\t',header=None)
classes_diabetes = pd.read_table('patients_classes.txt',sep='\t',header=None)
```

1. **Decision trees**
   - You can learn more about decision trees in Python here: `http://scikit-learn.org/stable/modules/tree.html`
   - Run the classifier to learn a model

     ```
     clf = tree.DecisionTreeClassifier()
     clf = clf.fit(data_diabetes, classes_diabetes)
     ```

   - Visualize the tree and save it as a .pdf

     ```
     feature_names = ['age', 'hba1c', 'insuline taken', 'other drugs taken']
     classes = ['DR','NDR']
     dot_data = tree.export_graphviz(clf, out_file=None,
                                     feature_names=feature_names,
                                     class_names=classes,
                                     filled=True, rounded=True,
                                     special_characters=True)
     graph = graphviz.Source(dot_data)
     graph.render("diabetes remission")
     ```

2. **Random forest**

   - You can learn more about the Random Forest in Python:
     `http://scikit-learn.org/stable/modules/generated/sklearn.ensemble.RandomForestClass`
     `html`
   - To estimate a model:

     ```
     clf = RandomForestClassifier(max_depth=2, random_state=0)
     clf.fit(data_diabetes, classes_diabetes)
     ```

   - To make prediction with the random forest:

     ```
     clf.predict(data_diabetes)
     ```

   - To plot the influence of each variable in the model:

     ```
     clf.feature_importances_
     ```

3. **Comparison with the state-of-the-art clinical score DiaRem**

   The DiaRem (Diabetes Remission score) was introduced recently by *Still et al.,2013* (see the references below), and can be summarized by the following table:

	Score
Age (years)	
<40	0
40–49	1
50–59	2
≥60	3
HbA_{1c} (%)	
<6·5%	0
6·5–6·9%	2
7·0–8·9%	4
≥9·0%	6
Other diabetes drugs	
No sulfonylureas or insulin-sensitising agent other than metformin	0
Sulfonylureas and insulin-sensitising agent other than metformin	3
Treatment with insulin	
No	0
Yes	10

Total score calculated by adding scores for each of the four variables.

Table 5: **Calculation of DiaRem score for prediction of the probability of diabetes remission after Roux-en-Y gastric bypass surgery**

   For a patient, if the sum of the scores over all clinical variables is < 7, we will classify this patient as one having the diabetes remission, otherwise, we will put him in the class of non-remission.

4. Compare the predictive power of the considered models (decision trees, random forest, and the DiaRem). What can you conclude?

References:

1. "The use of classification trees for bioinformatics"
   `http://moult.ibbr.umd.edu/JournalClubPresentations/Maya/Maya-04Feb2011-paper.pdf`
2. "Preoperative prediction of type 2 diabetes remission after Roux-en-Y gastric bypass surgery: a retrospective cohort study" `https://www.ncbi.nlm.nih.gov/pubmed/24579062`