Comparaci?n DT en R y en Python

October 22, 2018

1 Comparación DT en R y en Python

En este notebook compararé el comportamiento del modelo DecisionTree de R con el de Python, para ver si descrepan.

Iré modificando los parámetros que cada uno tiene, y veremos el accuracy que consigue cada uno

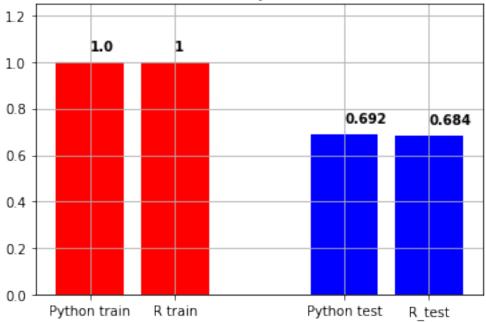
```
In [1]: sos_path = "../../datasets/falldetection/falldetection.csv"
```

1.1 En python3

```
In [2]: %use python3
        import pandas as pd
        import numpy as np
        from sklearn.tree import DecisionTreeClassifier
        import matplotlib.pyplot as plt
        df = pd.read_csv(sos_path)
In [3]: N = df.shape[0]
        prop_train = 2 / 3
        N_train = np.ceil(N * prop_train).astype(np.int64)
        N_{test} = N - N_{train}
In [4]: data = df.drop(["ACTIVITY"], 1)
        target = df.ACTIVITY
In [5]: data_train = data.iloc[:N_train]
        data_test = data.iloc[N_train:]
        target_train = target[:N_train]
        target_test = target[N_train:]
In [6]: dtc = DecisionTreeClassifier()
In [7]: dtc.fit(data_train, target_train)
```

```
Out[7]: DecisionTreeClassifier(class_weight=None, criterion='gini', max_depth=None,
                    max_features=None, max_leaf_nodes=None,
                    min_impurity_decrease=0.0, min_impurity_split=None,
                    min_samples_leaf=1, min_samples_split=2,
                    min_weight_fraction_leaf=0.0, presort=False, random_state=None,
                    splitter='best')
In [8]: train_score = dtc.score(data_train, target_train)
        test_score = dtc.score(data_test, target_test)
In [9]: train_score, test_score
Out [9]: (1.0, 0.6915750915750916)
1.2 En R
In [10]: %use R
         library(rpart)
         df = read.csv(sos_path)
In [11]: df $ACTIVITY = as.factor(df $ACTIVITY)
In [12]: N = nrow(df)
         prop_train = 2/3
         N_train = ceiling(N * prop_train)
         N_{\text{test}} = N - N_{\text{train}}
         data_train = df[1:N_train,]
         data_test = df[(N_train + 1):N,]
In [13]: ctrl = rpart.control(
             minsplit = 2,
             cp = 0,
             maxcomplete = 0,
             maxsurrogate = 0,
             usesurrogate = 0,
             xval = 10,
             surrogatestyle = 0,
             maxdepth = 30,
         )
In [14]: rpt = rpart(ACTIVITY ~ ., data = data_train,control = ctrl, method = "class")
In [15]: train_pred = predict(rpt, data_train, type = "class")
         test_pred = predict(rpt, data_test, type = "class")
In [16]: train_cf = table(data_train_$ACTIVITY, train_pred)
         test_cf = table(data_test$ACTIVITY, test_pred)
In [17]: R_train_score = sum(diag(train_cf)) / nrow(data_train)
         R_test_score = sum(diag(test_cf)) / nrow(data_test)
```

DT scores in Python and in R



1.3 Modificando min_samples_split

1.3.1 En R

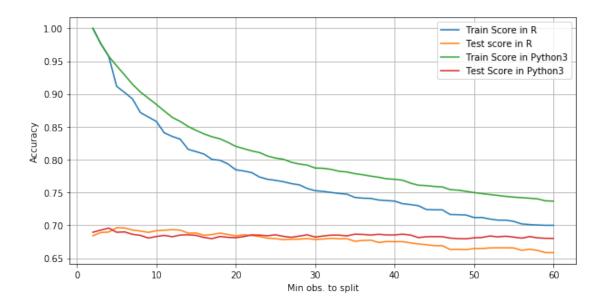
```
df $ ACTIVITY = as.factor(df $ ACTIVITY)
         N = nrow(df)
         prop_train = 2/3
         N_train = ceiling(N * prop_train)
         N_{test} = N - N_{train}
         data_train = df[1:N_train,]
         data_test = df[(N_train + 1):N,]
In [22]: spl = 2:60
In [23]: f = function(s)
                 ctrl = rpart.control(
                     minsplit = s,
                     cp = 0,
                     maxcomplete = 0,
                     maxsurrogate = 0,
                     usesurrogate = 0,
                     xval = 10,
                     surrogatestyle = 0,
                     maxdepth = 30,
                 rpt = rpart(ACTIVITY ~ ., data = data_train,control = ctrl, method = "class")
                 train_pred = predict(rpt, data_train, type = "class")
                 test_pred = predict(rpt, data_test, type = "class")
                 train_cf = table(data_train_$ACTIVITY, train_pred)
                 test_cf = table(data_test$ACTIVITY, test_pred)
                 R_train_score = sum(diag(train_cf)) / nrow(data_train)
                 R_test_score = sum(diag(test_cf)) / nrow(data_test)
                 return(list("train" = R_train_score, "test" = R_test_score))
             }
In [24]: res = sapply(spl, FUN = "f")
In [25]: R_train_scores = res[1, 1:ncol(res)]
         R_test_scores = res[2, 1:ncol(res)]
1.3.2 En Python3
In [26]: clf = DecisionTreeClassifier()
In [27]: python_train_scores = []
         python_test_scores = []
         for i in range(2,61):
             clf.set_params(min_samples_split = i)
             clf.fit(data_train, target_train)
```

```
train_score = clf.score(data_train, target_train)
    test_score = clf.score(data_test, target_test)

python_train_scores.append(train_score)
    python_test_scores.append(test_score)

In [28]: %get R_train_scores R_test_scores spl --from R
    plt.figure(figsize=(10,5))
    plt.plot(spl ,R_train_scores, label = "Train Score in R")
    plt.plot(spl, R_test_scores, label = "Test score in R")
    plt.plot(spl, python_train_scores, label = "Train Score in Python3")
    plt.plot(spl, python_test_scores, label = "Test Score in Python3")
    plt.grid(True)
    plt.legend()
    plt.xlabel("Min obs. to split")
    plt.ylabel("Accuracy")
    #plt.ylim(0.5,1.1)
```

Out[28]: Text(0, 0.5, 'Accuracy')



1.4 Modificando max_depth

```
N = nrow(df)
         prop_train = 2/3
         N_train = ceiling(N * prop_train)
         N_{test} = N - N_{train}
         data_train = df[1:N_train,]
         data_test = df[(N_train + 1):N,]
         d = 2:30
In [30]: f = function(d) {
             ctrl = rpart.control(
                     minsplit = 20,
                     cp = 0,
                     maxcomplete = 0,
                     maxsurrogate = 0,
                     usesurrogate = 0,
                     xval = 10,
                     surrogatestyle = 0,
                     maxdepth = d,
             rpt = rpart(ACTIVITY ~ ., data = data_train,control = ctrl, method = "class")
             train_pred = predict(rpt, data_train, type = "class")
             test_pred = predict(rpt, data_test, type = "class")
             train_cf = table(data_train_$ACTIVITY, train_pred)
             test_cf = table(data_test$ACTIVITY, test_pred)
             R_train_score = sum(diag(train_cf)) / nrow(data_train)
             R_test_score = sum(diag(test_cf)) / nrow(data_test)
             return(list("train" = R_train_score, "test" = R_test_score))
         }
In [31]: res = sapply(d, FUN = "f")
In [32]: R_train_scores = res[1, 1:ncol(res)]
         R_test_scores = res[2, 1:ncol(res)]
Nos pasamos a python
In [33]: clf = DecisionTreeClassifier()
In [34]: python_train_scores = []
         python_test_scores = []
         d = list(range(2,31))
         for i in d:
             clf.set_params(max_depth = i)
```

```
clf.fit(data_train, target_train)
             train_score = clf.score(data_train, target_train)
             test_score = clf.score(data_test, target_test)
            python_train_scores.append(train_score)
            python_test_scores.append(test_score)
In [35]: %get R_train_scores R_test_scores --from R
         plt.figure(figsize=(10,5))
         plt.plot(d ,R_train_scores, label = "Train Score in R")
         plt.plot(d, R_test_scores, label = "Test score in R")
         plt.plot(d, python_train_scores, label = "Train Score in Python3")
         plt.plot(d, python_test_scores, label = "Test Score in Python3")
         plt.grid(True)
        plt.legend()
         plt.xlabel("Max depth")
         plt.ylabel("Accuracy")
         #plt.ylim(0.5,1.1)
```

Out[35]: Text(0, 0.5, 'Accuracy')

