REGex

Task: 11

ID: SIRSS1196

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In [80]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

In [81]:

```
df=pd.read_csv("train.csv")
td=pd.read_csv("test.csv")
```

In [82]:

```
df.head(10)
```

Out[82]:

	id	species	margin1	margin2	margin3	margin4	margin5	margin6	margin7	margin8	margin9	margi
0	1	Acer_Opalus	0.007812	0.023438	0.023438	0.003906	0.011719	0.009766	0.027344	0.0	0.001953	0.033
1	2	Pterocarya_Stenoptera	0.005859	0.000000	0.031250	0.015625	0.025391	0.001953	0.019531	0.0	0.000000	0.007
2	3	Quercus_Hartwissiana	0.005859	0.009766	0.019531	0.007812	0.003906	0.005859	0.068359	0.0	0.000000	0.044
3	5	Tilia_Tomentosa	0.000000	0.003906	0.023438	0.005859	0.021484	0.019531	0.023438	0.0	0.013672	0.017
4	6	Quercus_Variabilis	0.005859	0.003906	0.048828	0.009766	0.013672	0.015625	0.005859	0.0	0.000000	0.005
5	8	Magnolia_Salicifolia	0.070312	0.093750	0.033203	0.001953	0.000000	0.152340	0.007812	0.0	0.003906	0.027
6	10	Quercus_Canariensis	0.021484	0.031250	0.017578	0.009766	0.001953	0.042969	0.039062	0.0	0.003906	0.019
7	11	Quercus_Rubra	0.000000	0.000000	0.037109	0.050781	0.003906	0.000000	0.003906	0.0	0.048828	0.003
8	14	Quercus_Brantii	0.005859	0.001953	0.033203	0.015625	0.001953	0.000000	0.023438	0.0	0.000000	0.021
9	15	Salix_Fragilis	0.000000	0.000000	0.009766	0.037109	0.072266	0.000000	0.000000	0.0	0.007812	0.001

10 rows × 194 columns

In [83]:

```
td.head(10)
```

Out[83]:

	id	margin1	margin2	margin3	margin4	margin5	margin6	margin7	margin8	margin9	margin10	margin11	margin1
0	4	0.019531	0.009766	0.078125	0.011719	0.003906	0.015625	0.005859	0.0	0.005859	0.023438	0.005859	0.02148
1	7	0.007812	0.005859	0.064453	0.009766	0.003906	0.013672	0.007812	0.0	0.033203	0.023438	0.009766	0.01953
2	9	0.000000	0.000000	0.001953	0.021484	0.041016	0.000000	0.023438	0.0	0.011719	0.005859	0.001953	0.02148
3	12	0.000000	0.000000	0.009766	0.011719	0.017578	0.000000	0.003906	0.0	0.003906	0.001953	0.000000	0.02929
4	13	0.001953	0.000000	0.015625	0.009766	0.039062	0.000000	0.009766	0.0	0.005859	0.000000	0.001953	0.03320
5	16	0.021484	0.033203	0.021484	0.009766	0.015625	0.035156	0.039062	0.0	0.003906	0.029297	0.013672	0.01953
6	19	0.015625	0.025391	0.046875	0.009766	0.005859	0.027344	0.042969	0.0	0.000000	0.027344	0.015625	0.00976
7	23	0.007812	0.031250	0.011719	0.050781	0.000000	0.117190	0.003906	0.0	0.011719	0.017578	0.056641	0.00195

	id	margin1	margin2	margin3	margin4	margin5	margin6	margin7	margin8	margin9	margin10	margin11	margin1
8	24	0.003906	0.007812	0.074219	0.017578	0.015625	0.003906	0.011719	0.0	0.009766	0.011719	0.023438	0.03710
9	28	0.000000	0.000000	0.005859	0.021484	0.054688	0.000000	0.015625	0.0	0.011719	0.005859	0.000000	0.03320

10 rows × 193 columns

In [84]:

```
df.info()
td.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 990 entries, 0 to 989
Columns: 194 entries, id to texture64
dtypes: float64(192), int64(1), object(1)
memory usage: 1.5+ MB
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 594 entries, 0 to 593
Columns: 193 entries, id to texture64
dtypes: float64(192), int64(1)
memory usage: 895.8 KB
```

In [85]:

```
print('train shape ',df.shape)
print('test shape ',td.shape)
```

```
train shape  (990, 194)
test shape  (594, 193)
```

In [86]:

```
df.columns
```

Out[86]:

```
Index(['id', 'species', 'margin1', 'margin2', 'margin3', 'margin4', 'margin5',
       'margin6', 'margin7', 'margin8',
       ...
       'texture55', 'texture56', 'texture57', 'texture58', 'texture59',
       'texture60', 'texture61', 'texture62', 'texture63', 'texture64'],
      dtype='object', length=194)
```

In [52]:

```
td.columns
```

Out[52]:

```
Index(['id', 'margin1', 'margin2', 'margin3', 'margin4', 'margin5', 'margin6',
       'margin7', 'margin8', 'margin9',
       ...
       'texture55', 'texture56', 'texture57', 'texture58', 'texture59',
       'texture60', 'texture61', 'texture62', 'texture63', 'texture64'],
      dtype='object', length=193)
```

In [53]:

```
df.nunique()
td.nunique()
```

Out[53]:

```
id           594
margin1       42
margin2       77
margin3       59
margin4       63
            ...
texture60     61
texture61     34
```

```
texture62    103
texture63     56
texture64     87
Length: 193, dtype: int64
```

In [54]:

```
print(df.isnull().sum())
print(td.isnull().sum())
```

```
id           0
species      0
margin1      0
margin2      0
margin3      0
            ..
texture60    0
texture61    0
texture62    0
texture63    0
texture64    0
Length: 194, dtype: int64
id           0
margin1      0
margin2      0
margin3      0
margin4      0
            ..
texture60    0
texture61    0
texture62    0
texture63    0
texture64    0
Length: 193, dtype: int64
```

In [55]:

```
from sklearn.preprocessing import LabelEncoder
enc=LabelEncoder()
df['species']=enc.fit_transform(df['species'])
```

In [56]:

```
X=df.drop(['id','species'],axis=1).values
Y=df[['species']].values
print(X.shape,Y.shape)
```

```
(990, 192) (990, 1)
```

In [57]:

```
td.head()
```

Out[57]:

	id	margin1	margin2	margin3	margin4	margin5	margin6	margin7	margin8	margin9	margin10	margin11	margin1
0	4	0.019531	0.009766	0.078125	0.011719	0.003906	0.015625	0.005859	0.0	0.005859	0.023438	0.005859	0.02148
1	7	0.007812	0.005859	0.064453	0.009766	0.003906	0.013672	0.007812	0.0	0.033203	0.023438	0.009766	0.01953
2	9	0.000000	0.000000	0.001953	0.021484	0.041016	0.000000	0.023438	0.0	0.011719	0.005859	0.001953	0.02148
3	12	0.000000	0.000000	0.009766	0.011719	0.017578	0.000000	0.003906	0.0	0.003906	0.001953	0.000000	0.02929
4	13	0.001953	0.000000	0.015625	0.009766	0.039062	0.000000	0.009766	0.0	0.005859	0.000000	0.001953	0.03320

5 rows × 193 columns

Decision Tree Classifier

In [58]:

```
from sklearn.tree import DecisionTreeClassifier
dtc=DecisionTreeClassifier(criterion='entropy',random_state=2)
dtc.fit(x_train,y_train)
```

Out[58]:

```
DecisionTreeClassifier(ccp_alpha=0.0, class_weight=None, criterion='entropy',
                       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='deprecated',
                       random_state=2, splitter='best')
```

In [92]:

```
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler().fit(X)
X = scaler.transform(X)
```

In [93]:

```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(X,Y,test_size=0.3,random_state=5)
```

Random Forest

In [94]:

```
from sklearn.ensemble import RandomForestClassifier
rf_classifier = RandomForestClassifier(n_estimators = 20,criterion = 'entropy', max_dept
h = 20, random_state = 5)
rf_classifier.fit(x_train, y_train)
```

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:3: DataConversionWarning: A
column-vector y was passed when a 1d array was expected. Please change the shape of y to
(n_samples,), for example using ravel().
  This is separate from the ipykernel package so we can avoid doing imports until
```

Out[94]:

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None,
                       criterion='entropy', max_depth=20, max_features='auto',
                       max_leaf_nodes=None, max_samples=None,
                       min_impurity_decrease=0.0, min_impurity_split=None,
                       min_samples_leaf=1, min_samples_split=2,
                       min_weight_fraction_leaf=0.0, n_estimators=20,
                       n_jobs=None, oob_score=False, random_state=5, verbose=0,
                       warm_start=False)
```

In [95]:

```
pred_train = rf_classifier.predict(x_train)
pred_test = rf_classifier.predict(x_test)
```

In [96]:

```
from sklearn.metrics import accuracy_score
print('Training Accuracy: ', accuracy_score(y_train, pred_train))
print('Testing Accuracy: ', accuracy_score(y_test, pred_test))
```

```
Training Accuracy:  1.0
Testing Accuracy:  0.8720538720538721
```

In [97]:

```
y_pred=rf_classifier.predict(x_test)
print(y_test,y_pred)
```

```
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```

In [98]:

```
print(accuracy_score(y_test,y_pred))
print(y_test,y_pred)
```

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 27 88  6 82 62 31 85 78 45  5 54 68 77 56 64 14  1  9 91 40 38 49  5 13
 79 44 55  4  1 56 63  7 33]
```

In [99]:

```
test_ids = df.pop('id')
x_test = df.values
```

In [100]:

```
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler().fit(X)
X = scaler.transform(X)
```

In [110]:

```
submission = pd.DataFrame(y_test)
```

In [106]:

```
submission.head(5)
```

Out[106]:

	Acer_Capillipes	Acer_Circinatum	Acer_Mono	Acer_Opalus	Acer_Palmatum	Acer_Pictum	Acer_Platanoids	Acer_Rubrum	…

	Acer_Capillipes	Acer_Circinatum	Acer_Mono	Acer_Opalus	Acer_Palmatum	Acer_Pictum	Acer_Platanoids	Acer_Rubrum
0	0.0	0.0	0.05	0.05	0.0	0.0	0.00	0.00
1	0.0	0.0	0.00	0.00	0.0	0.0	0.00	0.05
2	0.0	0.0	0.00	0.00	0.0	0.0	0.00	0.00
3	0.0	0.0	0.00	0.00	0.0	0.0	0.00	0.00
4	0.0	0.0	0.00	0.00	0.0	0.0	0.05	0.00

5 rows × 99 columns

In [71]:

```
submission.to_csv('submission_leaf_classification.csv')
```