Machine Learning Lab (PMCA507P)

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Random Forest Classifier

Collab url: https://colab.research.google.com/drive/1on_UwLL-DmtkESXxLLifEkB3qNEQ9clv?usp=sharing

Dataset url: https://www.kaggle.com/datasets/uciml/glass/data

Glass Classification Dataset

Import necessary libraries

```
import pandas as pd
from sklearn import tree
from sklearn.preprocessing import LabelEncoder
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier

df=pd.read_csv('/content/glass.csv')
```

df.head()

	RI	Na	Mg	Al	Si	K	Ca	Ва	Fe	Туре	\blacksquare
0	1.52101	13.64	4.49	1.10	71.78	0.06	8.75	0.0	0.0	1	ılı
1	1.51761	13.89	3.60	1.36	72.73	0.48	7.83	0.0	0.0	1	
2	1.51618	13.53	3.55	1.54	72.99	0.39	7.78	0.0	0.0	1	
3	1.51766	13.21	3.69	1.29	72.61	0.57	8.22	0.0	0.0	1	
4	1.51742	13.27	3.62	1.24	73.08	0.55	8.07	0.0	0.0	1	

```
View recommended plots
              Generate code with df
 Next steps:
df.rename(columns={
    'RI': 'refractive_index',
    'Na': 'Sodium',
    'Mg': 'Magnesium',
    'Al': 'Aluminum',
    'Si': 'Silicon',
    'K': 'Potassium',
    'Ca': 'Calcium',
    'Ba': 'Barium',
    'Fe': 'Iron'
}, inplace=True)
df.head()
                                                                                                           refractive_index Sodium Magnesium Aluminum Silicon Potassium Calcium Barium Iron Type
      0
                  1.52101
                            13.64
                                         4.49
                                                   1.10
                                                           71.78
                                                                       0.06
                                                                                8.75
                                                                                         0.0
                                                                                               0.0
                                                                                                           th
      1
                  1.51761
                            13.89
                                         3.60
                                                   1.36
                                                           72.73
                                                                                7.83
                                                                                               0.0
                                                                                                      1
                                                                       0.48
                                                                                         0.0
      2
                  1.51618
                            13.53
                                         3.55
                                                   1.54
                                                           72.99
                                                                       0.39
                                                                                7.78
                                                                                         0.0
                                                                                               0.0
                                                                                                      1
      3
                  1.51766
                            13.21
                                         3.69
                                                   1.29
                                                           72.61
                                                                                8.22
                                                                                               0.0
                                                                                                      1
                                                                       0.57
                                                                                         0.0
      4
                  1.51742
                            13.27
                                         3.62
                                                   1.24
                                                           73.08
                                                                       0.55
                                                                                8.07
                                                                                         0.0
                                                                                               0.0
                                                                                                      1
                                      View recommended plots
              Generate code with df
 Next steps:
df.describe()
```

	refractive_index	Sodium	Magnesium	Aluminum	Silicon	Potassium	Calcium	Barium
count	214.000000	214.000000	214.000000	214.000000	214.000000	214.000000	214.000000	214.000000
mean	1.518365	13.407850	2.684533	1.444907	72.650935	0.497056	8.956963	0.175047
std	0.003037	0.816604	1.442408	0.499270	0.774546	0.652192	1.423153	0.497219
min	1.511150	10.730000	0.000000	0.290000	69.810000	0.000000	5.430000	0.000000
25%	1.516522	12.907500	2.115000	1.190000	72.280000	0.122500	8.240000	0.000000
50%	1.517680	13.300000	3.480000	1.360000	72.790000	0.555000	8.600000	0.000000
75%	1.519157	13.825000	3.600000	1.630000	73.087500	0.610000	9.172500	0.000000
max	1.533930	17.380000	4.490000	3.500000	75.410000	6.210000	16.190000	3.150000
1								,

Encode categorical variables using LabelEncoder

```
label_encoder = LabelEncoder()
for col in df.columns:
   df[col] = label_encoder.fit_transform(df[col])
```

Split the data into features (X) and target (y)

```
X = df.drop('Type', axis=1)
y = df['Type']
```

Split the data into training and testing sets

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

Create a Decision tree classifier

```
dt_clf = DecisionTreeClassifier()
dt_clf.fit(X_train, y_train)

v DecisionTreeClassifier
DecisionTreeClassifier()
```

Print decision tree classifier accuracy

```
dt_pred = dt_clf.predict(X_test)
dt_accuracy = accuracy_score(y_test, dt_pred)
print("Decision Tree Classifier Accuracy:", dt_accuracy)

Decision Tree Classifier Accuracy: 0.7209302325581395
```

Create Radom Forest Classifier model

```
RandomForestClassifier
RandomForestClassifier(criterion='entropy', n_estimators=44)
```

Predict and evaluate the random forest classifier

```
rf_pred = rf_clf.predict(X_test)
rf_accuracy = accuracy_score(y_test, rf_pred)
print("Random Forest Classifier Accuracy:", rf_accuracy)

Random Forest Classifier Accuracy: 0.813953488372093
```

Print classification report and confusion matrix

```
print("Classification Report:")
print(classification_report(y_test, rf_pred))
print("Confusion Matrix:")
print(confusion_matrix(y_test, rf_pred))
    Classification Report:
                 precision
                             recall f1-score
                                              support
              0
                      0.71
                               1.00
                                        0.83
                                                   10
                      0.79
                                        0.76
                                                   15
              1
                               0.73
              2
                      1.00
                               0.67
                                        0.80
                                                    3
                      0.75
                               1.00
                                        0.86
                      1.00
                               0.67
                                        0.80
                                                    3
                      1.00
                               0.78
                                        0.88
                                        0.81
                                                   43
        accuracy
       macro avg
                      0.88
                               0.81
                                        0.82
                                                   43
    weighted avg
                      0.84
                               0.81
                                        0.81
                                                   43
    Confusion Matrix:
    [[10 0 0 0 0
     [311 0 1 0 0]
     [102000]
     [000300]
     [0 1 0 0 2 0]
```

[020007]]

Plot the first decision tree in the random forest

```
plt.figure(figsize=(15, 5))
tree.plot_tree(rf_clf.estimators_[0])
plt.show()
```

