Dear Professor: Mr.Manthouri

Produced By: Ghasemi, morteza

https://github.com/Morteza-Ghasemi

January 2021

- Classification is a large domain in the field of statistics and machine learning. Generally, classification can be broken down into two areas
 - Binary classification, where we wish to group an outcome into one of two groups.
 - Multi-class classification, where we wish to group an outcome into one of multiple (more than two) groups.

- Logistic Regression is a type of Generalized Linear Model (GLM).
- uses a logistic function to model a binary variable based on any kind of independent variables.
- LR = LogisticRegression()
- LR. fit(X, y)

- The task is to predict whether a bank currency note is authentic or not based upon four attributes of the note i.e. skewness of the wavelet transformed image, variance of the image, entropy of the image, and curtosis of the image.
- This is a binary classification problem and then will use **Logistic Regression** algorithm to solve this problem.

Implementing Logistic Regression with Scikit-Learn

Importing libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
```

- Importing the Dataset
 - ✓ The data is available for download at the following link:

 https://drive.google.com/file/d/13nw-uRXPY8XIZQxKRNZ3yYlho-CYm_Qt/view
 - ✓ The detailed information about the data is available at the following link: https://archive.ics.uci.edu/ml/datasets/banknote+authentication

Hints:

• the dataset is banknote authentication

Abstract: Data were extracted from images that were taken for the evaluation of an authentication procedure for banknotes.

Data Set Characteristics:	Multivariate	Number of Instances:	1372	Area:	Computer
Attribute Characteristics:	Real	Number of Attributes:	5	Date Donated	2013-04-16
Associated Tasks:	Classification	Missing Values?	N/A	Number of Web Hits:	290722

- Exploratory Data Analysis
 - ✓ The following code reads bank currency note data into pandas dataframe:

 bankdata = pd.read_csv(r"D:\...\bill_authentication.csv")
 - ✓ To see the rows and columns and of the data, execute the following command: bankdata.shape

Implementing Logistic Regression with Scikit-Learn

Exploratory Data Analysis

- ✓ To get a feel of how our dataset actually looks, execute the following command: bankdata.head()
- ✓ You can see that all of the attributes in the dataset are numeric. The label is also numeric i.e. 0 and 1.

- Data Preprocessing
 - ✓ Dividing the data into attributes and labels :
 - \checkmark X = bankdata.drop('Class', axis=1)
 - \checkmark y = bankdata['Class']
 - ✓ dividing the data into training and testing sets:
 - ✓ from sklearn.model_selection import train_test_split
 - ✓ X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.20)

- Training the Algorithm
 - ✓ The fit method of **Logistic Regression** class is called to train the algorithm on the training data, which is passed as a parameter to the fit method.
 - ✓ from sklearn.linear_model import LogisticRegression
 - ✓ logreg = LogisticRegression()
 - ✓ logreg.fit(X_train, y_train)

- Making Predictions
 - ✓ To make predictions, the predict method of the **Logistic Regression** class is used.
 - ✓ y_pred = logreg.predict(X_test)

- Evaluating the Algorithm
 - ✓ Confusion matrix, precision, recall, and F1 measures are the most commonly used metrics for classification tasks.
 - ✓ from sklearn.metrics import classification_report, confusion_matrix
 - ✓ print(confusion_matrix(y_test,y_pred))
 - ✓ print(classification_report(y_test,y_pred))

- Evaluating the Algorithm(Penalty = None)
 - ✓ The output of the Logistic Regression(penalty='none', solver='saga') looks like this:

```
In [18]: print
[[146 2]
[ 2 125]]
In [19]: print
            precision
                         recall f1-score
                                            support
                  0.99
                           0.99
                                     0.99
                                                 148
                 0.98
                           0.98
                                     0.98
                                                 127
avg / total
                 0.99
                                     0.99
                                                275
                            0.99
```

