ML with sklearn

April 14, 2023

```
[3]: # Name: Emmanuel Asante
    # Class: CS 4375
    # Assg: Sklearn Python
    import pandas as pd
    print("Machine Learning with SKLearn - Emmanuel

→Asante\n-----\n")
    # 1a) Reading csv - file_name auto.csv
    try:
        print("QUESTION 1:\n----")
        print("1a) Attempting to Open the file: <Auto.csv>")
        data_frame = pd.read_csv("C:\\Users\\Manny\\Box\\University\\Sem 5 (Spring_
     →2023)\\CS 4375\\Assignments\\SK Learn\\Auto.csv")
        print("File Opened\n")
    except:
        print("1b) Outputting the first few rows pf the data")
    # 1b) Outputting the first few rows
    print("1b) Attempting to Open the file: <Auto.csv>")
    print(data_frame.head(), "\n")
    data_frame.shape
    # 1c) Dimensions of the data
    print("1c) Dimensions of the data:\n Rows: {} \n Cols: {} ".

→format(len(data_frame), len(data_frame.axes[1])))
    Machine Learning with SKLearn - Emmanuel Asante
    QUESTION 1:
    -----
    1a) Attempting to Open the file: <Auto.csv>
    File Opened
    1b) Attempting to Open the file: <Auto.csv>
       mpg cylinders displacement horsepower weight acceleration year
```

```
0 18.0
                     8
                               307.0
                                             130
                                                    3504
                                                                  12.0 70.0 \
    1 15.0
                     8
                               350.0
                                             165
                                                    3693
                                                                  11.5 70.0
    2 18.0
                                                                  11.0 70.0
                     8
                               318.0
                                             150
                                                    3436
    3 16.0
                     8
                               304.0
                                             150
                                                    3433
                                                                  12.0 70.0
    4 17.0
                     8
                                                                   NaN 70.0
                               302.0
                                             140
                                                    3449
       origin
                                    name
               chevrolet chevelle malibu
    0
    1
            1
                       buick skylark 320
    2
                      plymouth satellite
            1
    3
                           amc rebel sst
            1
    4
            1
                             ford torino
    1c) Dimensions of the data:
        Rows: 392
        Cols: 9
[4]: # Question 2
    print("\nQUESTION 2:\n----\n")
    print("2a) Describing mpg, weight, and year columns \n")
    print (data_frame[ ['mpg', 'weight', 'year']].describe())
    print("\n2b) Summary of Description")
    print("MPG: Range = [{}, {}] Average = {:.2f}".format(data_frame['mpg'].

min(), data_frame['mpg'].max(), data_frame['mpg'].mean()))
```

-format(data_frame['cylinders'].min(), data_frame['cylinders'].max(),u

print("Year: Range = [{}, {}] Average = {:.2f}".format(data_frame['year'].

¬min(), data_frame['year'].max(), data_frame['year'].mean()))

Average = $\{:.2f\}$ ".

QUESTION 2:

2a) Describing mpg, weight, and year columns

print("Cyl: Range = [{}, {}]

¬data_frame['cylinders'].mean()))

	mpg	weight	year
count	392.000000	392.000000	390.000000
mean	23.445918	2977.584184	76.010256
std	7.805007	849.402560	3.668093
min	9.000000	1613.000000	70.000000
25%	17.000000	2225.250000	73.000000
50%	22.750000	2803.500000	76.000000
75%	29.000000	3614.750000	79.000000
max	46.600000	5140.000000	82.000000

```
2b) Summary of Description
    MPG: Range = [9.0, 46.6] Average = 23.45
    Cyl: Range = [3, 8]
                               Average = 5.47
    Year: Range = [70.0, 82.0] Average = 76.01
[5]: # Question 3
     print("\nQUESTION 3:\n----\n")
     print("3a) Data Types of each column")
     print (data_frame.dtypes, "\n")
     print("3b) Changing cylinder to categorical")
     data_frame['origin'] = data_frame['origin'].astype('category')
     print ("changed")
     print("3c) Changing cylinder to categorical witouht cat.codes")
     data_frame['cylinders'] = data_frame['cylinders'].astype('category')
     print ("changed")
     print (data_frame.dtypes)
    QUESTION 3:
    3a) Data Types of each column
                    float64
    mpg
                      int64
    cylinders
    displacement
                    float64
                      int64
    horsepower
    weight
                      int64
    acceleration
                    float64
    year
                    float64
                      int64
    origin
    name
                     object
    dtype: object
    3b) Changing cylinder to categorical
    changed
    3c) Changing cylinder to categorical witouht cat.codes
    changed
                     float64
    mpg
    cylinders
                    category
    displacement
                     float64
    horsepower
                       int64
    weight
                       int64
    acceleration
                     float64
                     float64
    year
    origin
                    category
```

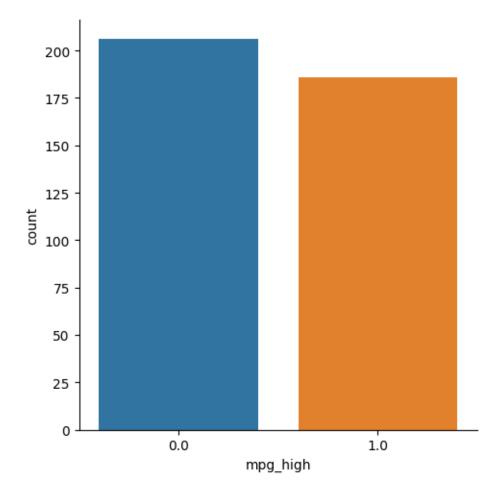
```
object
    name
    dtype: object
[6]: # Question 4
    print("\nQUESTION 4:\n----\n")
    print("4a) Dealing wiht NA's")
    # data_frame.dropna(inplace= True)
    # print("4b) New dimensions of the data:\n Rows: {} \n Cols: {} ".

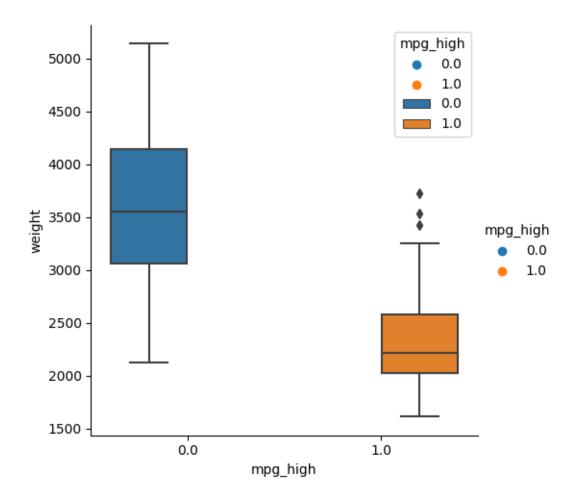
→format(len(new_data_frame), len(data_frame.axes[1])))
    QUESTION 4:
    _____
    4a) Dealing wiht NA's
[7]: # Question 5
    print("\nQUESTION 5:\n----\n")
    print("5a) Creating a new column mpg_high which is a categorical version of ⊔
      ympg")
    pd.set_option('display.max_rows', 10)
    mpg_high = data_frame['mpg'].copy()
    mpg_avg = mpg_high.mean()
    print (mpg_avg)
    for i in range(0, len(mpg_high)):
        val = mpg_high.iloc[i]
        if val > mpg_avg:
            mpg_high.at[i] = 1
        else:
            mpg_high.at[i] = 0
    mpg_high = mpg_high.astype('category')
    data_frame['mpg_high'] = mpg_high
    data_frame
    print("5b) Removing mpg and name")
    data_frame = data_frame.drop(['mpg', 'name'], axis = 1)
    data_frame
    QUESTION 5:
    5a) Creating a new column mpg_high which is a categorical version of mpg
    23.445918367346938
```

5b) Removing mpg and name

```
[7]:
         cylinders
                   displacement horsepower weight acceleration year origin
                           307.0
                                                               12.0 70.0
    0
                 8
                                         130
                                                3504
                                                                               1
                                                                                 \
                 8
     1
                           350.0
                                         165
                                                3693
                                                               11.5 70.0
                                                                               1
     2
                 8
                           318.0
                                         150
                                                3436
                                                               11.0 70.0
                                                                               1
     3
                 8
                                                               12.0 70.0
                           304.0
                                         150
                                                3433
                                                                               1
     4
                 8
                           302.0
                                         140
                                                3449
                                                               NaN 70.0
     . .
                                                                •••
                                                               15.6 82.0
     387
                 4
                           140.0
                                          86
                                                2790
                                                                               1
     388
                            97.0
                                                2130
                                                               24.6 82.0
                                                                               2
                 4
                                          52
     389
                 4
                           135.0
                                          84
                                                2295
                                                              11.6 82.0
                                                                               1
     390
                 4
                           120.0
                                          79
                                                2625
                                                              18.6 82.0
                                                                               1
     391
                 4
                           119.0
                                          82
                                                2720
                                                              19.4 82.0
                                                                               1
         mpg_high
     0
              0.0
              0.0
     1
     2
              0.0
     3
              0.0
     4
              0.0
     387
              1.0
     388
              1.0
     389
              1.0
     390
              1.0
     391
              1.0
     [392 rows x 8 columns]
[8]: # Question 6
     print("\nQUESTION 6:\n----\n")
     import seaborn as sns
     print("6a)Seaborn catplot on mpg_high")
     sns.catplot(x = 'mpg_high', data=data_frame, kind='count')
     print("6b)Seaborn catplot on mpg_high")
     sns.relplot(x='horsepower', y = 'weight', data=data_frame, hue='mpg_high')
     print("6c)Seaborn boxplot on mpg_high")
     sns.boxplot(x='mpg_high', y = 'weight', data=data_frame, hue='mpg_high')
    QUESTION 6:
    6a)Seaborn catplot on mpg_high
    6b)Seaborn catplot on mpg_high
    6c)Seaborn boxplot on mpg_high
```

[8]: <Axes: xlabel='mpg_high', ylabel='weight'>





```
print("\n7c) Train and Test Data with mpg_high removed\n")
print("Train Data")
print("-----\n")
print(train_data)

print("Test Data")
print("-----\n")
print(test_data)

print("\n7d) Dimensions of the Test and Train Data\n")
print(" - Dimensions of Autos Data Frame:",data_frame.shape)
print(" - Dimensions of Train Data Frame:",train_data.shape)
print(" - Dimensions of Test Data Frame:",test_data.shape)
```

QUESTION 7:

7a) Dividing Data set in test and train data

- Train Size: 313 - Test Size: 79

7b) Setting Seed Ransom seed: 1234

7c) Train and Test Data with mpg_high removed

Train Data

	cylinders	displacement	horsepower	weight	acceleration	year	origin
0	8	307.0	130	3504	12.0	70.0	1
1	8	350.0	165	3693	11.5	70.0	1
2	8	318.0	150	3436	11.0	70.0	1
3	8	304.0	150	3433	12.0	70.0	1
4	8	302.0	140	3449	NaN	70.0	1
	•••	•••					
308	4	89.0	60	1968	18.8	80.0	3
309	4	98.0	70	2120	15.5	80.0	1
310	4	86.0	65	2019	16.4	80.0	3
311	4	151.0	90	2678	16.5	80.0	1
312	4	140.0	88	2870	18.1	80.0	1

[313 rows x 7 columns]

Test Data

```
displacement horsepower weight acceleration year origin
    cylinders
313
            4
                      151.0
                                     90
                                            3003
                                                          20.1 80.0
314
            6
                      225.0
                                     90
                                            3381
                                                          18.7 80.0
                                                                          1
            4
                       97.0
                                                          15.8 80.0
                                                                          2
315
                                     78
                                            2188
316
                      134.0
                                            2711
                                                          15.5 80.0
                                                                          3
                                     90
317
                      120.0
                                     75
                                            2542
                                                          17.5 80.0
                                                                          3
. .
                                      •••
                                                            •••
                      140.0
                                            2790
                                                          15.6 82.0
387
            4
                                     86
                                                                          1
                                                          24.6 82.0
388
            4
                       97.0
                                     52
                                           2130
                                                                          2
            4
                      135.0
                                            2295
                                                          11.6 82.0
389
                                     84
                                                                          1
            4
                      120.0
                                     79
                                            2625
                                                          18.6 82.0
390
                                                                          1
                                                          19.4 82.0
391
            4
                      119.0
                                     82
                                            2720
                                                                          1
```

[79 rows x 7 columns]

- 7d) Dimensions of the Test and Train Data
 - Dimensions of Autos Data Frame: (392, 8)
 - Dimensions of Train Data Frame: (313, 7)
 - Dimensions of Test Data Frame: (79, 7)

```
[10]: # Question 8
    print("\nQUESTION 8:\n----\n")
    from sklearn.linear_model import LogisticRegression
    from sklearn.model_selection import train_test_split

print("8a) Dividing Data set in test and train data")

# fit the model with data
Y = 'horsepower'

data_frame.drop (Y, axis=1)

X_train, X_test, Y_train, Y_test = train_test_split(data_frame.drop(Y, axis=1),u-data_frame[Y], test_size = 0.2)
    print(X_train)
    print(Y_train)

LogReg = LogisticRegression()
# LogReg.fit(X_train, Y_train)
```

QUESTION 8:

8a) Dividing Data set in test and train data cylinders displacement weight acceleration year origin mpg_high

```
116
            4
                        68.0
                                1867
                                               19.5 73.0
                                                               2
                                                                       1.0
248
            8
                       318.0
                                3735
                                               13.2 78.0
                                                               1
                                                                       0.0
                       250.0
                                3302
                                               15.5 71.0
                                                                       0.0
35
            6
                                                               1
245
            4
                        85.0
                                2070
                                               18.6 78.0
                                                               3
                                                                       1.0
146
            4
                       116.0
                                2246
                                               14.0 74.0
                                                               2
                                                                       1.0
. .
                                                               3
                                                                       1.0
                       91.0
                                               16.0 81.0
345
            4
                                1985
                       119.0
                                               15.0 80.0
                                                                       1.0
318
                                2434
                                                               3
            4
                       225.0
33
            6
                                3439
                                               15.5 71.0
                                                               1
                                                                       0.0
88
                       318.0
                                               12.5 73.0
                                                                       0.0
            8
                                3777
                                                               1
8
            8
                       455.0
                                4425
                                               10.0 70.0
                                                               1
                                                                       0.0
```

[313 rows x 7 columns]

Name: horsepower, Length: 313, dtype: int64

[]: