NEURAL NETWORK DEEP LEARNING

ICP 3

700758238

AKHIL KASANAGOTTU

GitHub:

Repository URL for the source code:
https://github.com/axk82380/NNPL/tree/main/ICP_3

Question 1: Data Manipulation

a. Read the provided CSV file 'data.csv'.

- Read_csv(*args) is used to read csv file.
- Info() is used to print type of variables

```
import numpy as np
import pandas as pd
```

```
[26] #Read the provided CSV file 'data.csv'
     data_Manip = pd.read_csv("/content/sample_data/data.csv")
     data_Manip.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 169 entries, 0 to 168
Data columns (total 4 columns):
 #   Column    Non-Null Count  Dtype
---  ------    --------------  -----
 0   Duration  169 non-null    int64
 1   Pulse     169 non-null    int64
 2   Maxpulse  169 non-null    int64
 3   Calories  164 non-null    float64
dtypes: float64(1), int64(3)
memory usage: 5.4 KB
```

c. Show the basic statistical description about the data.

- head() is used to get the first n rows.

```
[7] #(c) Show the basic statistical description about the data.
    data_Manip.head()
```

	Duration	Pulse	Maxpulse	Calories
0	60	110	130	409.1
1	60	117	145	479.0
2	60	103	135	340.0
3	45	109	175	282.4
4	45	117	148	406.0

d(i). Replace the null values with the mean.

- Filling the null values with mean of given data.

```
[9] data_Manip.fillna(data_Manip.mean(), inplace=True)
    data_Manip.isnull().any()
```

	0
Duration	False
Pulse	False
Maxpulse	False
Calories	False

dtype: bool

```
[10] #d(i)Replace the null values with the mean
     column_means = data_Manip.mean()
     print(column_means)
     data_Manip = data_Manip. fillna(column_means)
     print(data_Manip.head(20))
```

```
Duration     63.846154
Pulse       107.461538
Maxpulse    134.047337
Calories    375.790244
dtype: float64
    Duration  Pulse  Maxpulse    Calories
0         60    110       130  409.100000
1         60    117       145  479.000000
2         60    103       135  340.000000
3         45    109       175  282.400000
4         45    117       148  406.000000
5         60    102       127  300.000000
6         60    110       136  374.000000
7         45    104       134  253.300000
8         30    109       133  195.100000
9         60     98       124  269.000000
10        60    103       147  329.300000
11        60    100       120  250.700000
12        60    106       128  345.300000
13        60    104       132  379.300000
14        60     98       123  275.000000
15        60     98       120  215.200000
16        60    100       120  300.000000
17        45     90       112  375.790244
18        60    103       123  323.000000
19        45     97       125  243.000000
```

e. Select at least two columns and aggregate the data using: min, max, count, mean.

```
[11] #(e)Select at least two columns and aggregate the data using: min, max, count, mean.
     res = data_Manip.agg({'Calories': ['mean', 'min','max', 'count'],'Pulse': ['mean', 'min', 'max', 'count']})
     print(res)
```

```
          Calories       Pulse
mean    375.790244  107.461538
min      50.300000   80.000000
max    1860.400000  159.000000
count   169.000000  169.000000
```

f. Filter the data frame to select the rows with calories values between 500 and 1000.

```
[12] #(f)Filter the dataframe to select the rows with calories values between 500 and 1000
     filter_data_Manip1=data_Manip[(data_Manip['Calories'] > 500) & (data_Manip['Calories'] < 1000)]
     print(filter_data_Manip1)
```

```
     Duration  Pulse  Maxpulse  Calories
51         80    123       146     643.1
62        160    109       135     853.0
65        180     90       130     800.4
66        150    105       135     873.4
67        150    107       130     816.0
72         90    100       127     700.0
73        150     97       127     953.2
75         90     98       125     563.2
78        120    100       130     500.4
90        180    101       127     600.1
99         90     93       124     604.1
103        90     90       100     500.4
106       180     90       120     800.3
108        90     90       120     500.3
```

g. Filter the data frame to select the rows with calories values > 500 and pulse < 100.

```
[13] #(g)Filter the dataframe to select the rows with calories values > 500 and pulse < 100.
     filter_data_Manip2=data_Manip[(data_Manip['Calories'] > 500) & (data_Manip['Pulse'] < 100)]
     print(filter_data_Manip2)
```

```
     Duration  Pulse  Maxpulse  Calories
65        180     90       130     800.4
70        150     97       129    1115.0
73        150     97       127     953.2
75         90     98       125     563.2
99         90     93       124     604.1
103        90     90       100     500.4
106       180     90       120     800.3
108        90     90       120     500.3
```

h. Create a new “df_modified” data frame that contains all the columns from df except for

“Maxpulse”.

```
[14] #(h)Create a new “df_modified” dataframe that contains all the columns from dst_data except for
     #“Maxpulse”.
     data_modified = data_Manip.loc[:, data_Manip.columns != 'Maxpulse']
     print(data_modified)
```

```
     Duration  Pulse  Calories
0          60    110     409.1
1          60    117     479.0
2          60    103     340.0
3          45    109     282.4
4          45    117     406.0
..        ...    ...       ...
164        60    105     290.8
165        60    110     300.0
166        60    115     310.2
167        75    120     320.4
168        75    125     330.4

[169 rows x 3 columns]
```

i. Delete the “Maxpulse” column from the main df dataframe

- drop() is used to delete a column in data

```
[15] #(i). Delete the “Maxpulse” column from the main dst_data dataframe
     data_Manip.drop('Maxpulse', inplace=True, axis=1)
     print(data_Manip.dtypes)
```

```
Duration      int64
Pulse         int64
Calories    float64
dtype: object
```

j. Convert the datatype of Calories column to int datatype.

- Converting from float to int using astype(*args) function.

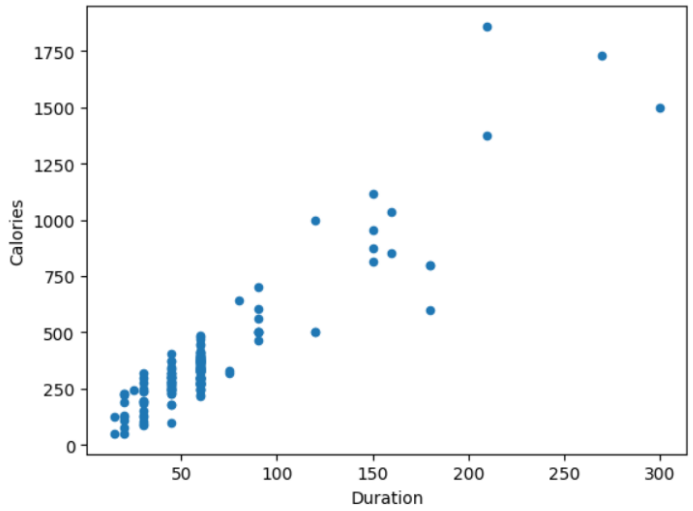
```
[16] #(j). Convert the datatype of Calories column to int datatype
     data_Manip["Calories"] = data_Manip["Calories"].astype(float).astype(int)
     print(data_Manip.dtypes)
```

```
Duration    int64
Pulse       int64
Calories    int64
dtype: object
```

k. Using pandas create a scatter plot for the two columns (Duration and Calories).

```
[17] #(k)Using pandas create a scatter plot for the two columns (Duration and Calories).
     sp1 = data_Manip.plot.scatter(x='Duration',y='Calories')
     print(sp1)
```

```
Axes(0.125,0.11;0.775x0.77)
```



Question 2. Linear Regression

a) Import the given “Salary_Data.csv”

- read_csv(*args) function is used to read data.

```
[19] # 2(a) Import the given "Salary_Data.csv"
     Lin_Re = pd.read_csv('/content/sample_data/Salary_Data.csv')
     Lin_Re.info()
     Lin_Re.head()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 30 entries, 0 to 29
Data columns (total 2 columns):
 #   Column           Non-Null Count  Dtype
---  ------           --------------  -----
 0   YearsExperience  30 non-null     float64
 1   Salary           30 non-null     float64
dtypes: float64(2)
memory usage: 608.0 bytes
```

	YearsExperience	Salary
0	1.1	39343.0
1	1.3	46205.0
2	1.5	37731.0
3	2.0	43525.0
4	2.2	39891.0

b) Split the data in train_test partitions, such that 1/3 of the data is reserved as test subset.

```
[20] #excluding last column i.e., years of experience column
     A = Lin_Re.iloc[:, :-1].values
     #only salary column
     B = Lin_Re.iloc[:, 1].values
```

```
[21] # (b) Split the data in train_test partitions, such that 1/3 of the data is reserved as test subset.
     from sklearn.model_selection import train_test_split
     A_train, A_test, B_train, B_test = train_test_split(A, B, test_size=1/3, random_state=0)
```

c) Train and predict the model.

- Training and predicting data using LinearRegression() model.

```
[22] # (c) Train and predict the model.
     from sklearn.linear_model import LinearRegression
     lRegression = LinearRegression()
     lRegression.fit(A_train, B_train)
     B_Pred = lRegression.predict(A_test)
     B_Pred
```

```
array([ 40835.10590871, 123079.39940819,  65134.55626083,  63265.36777221,
       115602.64545369, 108125.8914992 , 116537.23969801,  64199.96201652,
        76349.68719258, 100649.1375447 ])
```

d) Calculate the mean_squared error

- Ratio of sum_error to size of test is mean_squared error.

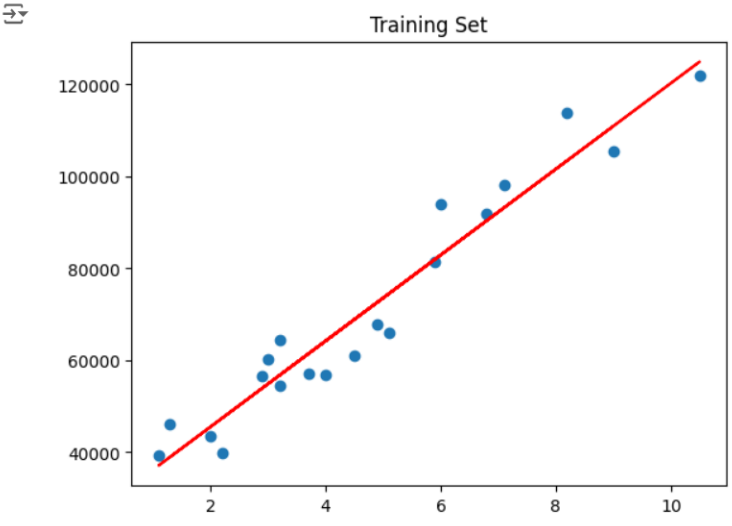
```
[23] # (d) Calculate the mean_squared error
     Sum_Serror = np.sum((B_Pred - B_test) ** 2)
     mean_squared_error = Sum_Serror / B_test.size
     mean_squared_error
```

```
21026037.329511296
```

e) Visualize both train and test data using scatter plot.

- Plotting using LinearRegression() and printing using show() function.

```
[24] # (e) Visualize both train and test data using scatter plot.
     import matplotlib.pyplot as plt
     # Training Data set
     plt.scatter(A_train, B_train)
     plt.plot(A_train, lRegression.predict(A_train), color='red')
     plt.title('Training Set')
     plt.show()
```



- For Testing data set

```
[25] # Testing Data set
     plt.scatter(A_test, B_test)
     plt.plot(A_test, lRegression.predict(A_test), color='red')
     plt.title('Testing Set')
     plt.show()
```