

Name: Krishna Bhatu More

Class: BTECH

Year: Second

Branch: Computer

College: R.C.Patel Institute Of Technology, Shirpur

Program: Data Science(Nov-22)

Major Project-1: Chose dataset of your choice and apply regressor

Major Project-2: Choose any dataset of your choice and perform EDA for atleast 15 different facts

Major Project-1: Years vs Salary Model using regressor

Major Project 1-Regression.ipynb ☆

File Edit View Insert Runtime Tools Help Last saved at January 9

Comment Share Settings User

Code + Text Connect Editing

```
##MAJOR PROJECT##
#MODEL USING LINEAR REGRESSION#
#YEARSEXPERIENCE VS SALARY
#1.Take the data and create dataframe
import pandas as pd
df = pd.read_csv('/content/Salary_Data.csv')
df
#CREATE A SIMPLE MODEL,WHICH COULD PREDICT SALARY BASED ON YEARSEXPERIENCE USING LINEAR REGRESSION.
```

	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
5	2.9	56642.0
6	3.0	60150.0

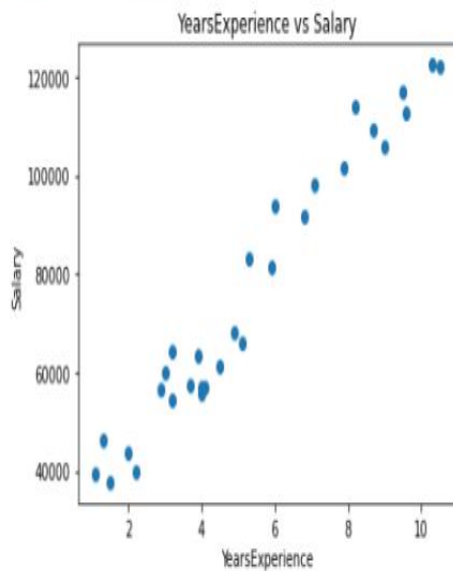
7	3.2	54445.0
8	3.2	64445.0
9	3.7	57189.0
10	3.9	63218.0
11	4.0	55794.0
12	4.0	56957.0
13	4.1	57081.0
14	4.5	61111.0
15	4.9	67938.0
16	5.1	66029.0
17	5.3	83088.0
18	5.9	81363.0
19	6.0	93940.0
20	6.8	91738.0

21	7.1	98273.0
22	7.9	101302.0
23	8.2	113812.0
24	8.7	109431.0
25	9.0	105582.0
26	9.5	116969.0
27	9.6	112635.0
28	10.3	122391.0
29	10.5	121872.0



```
[ ] #2.We are not performing step no 2 ,for this dataset
#3.DATA VISUALISATION - CREATION of GRAPHS
import matplotlib.pyplot as plt
#plt.scatter(x-axis,y-axis)
plt.scatter(df['YearsExperience'],df['Salary'])
plt.title('YearsExperience vs Salary')
plt.xlabel('YearsExperience')
[ ] plt.ylabel('Salary')
```

Text(0, 0.5, 'Salary')



```
[ ] #4.
#INPUT(x) -YearsExperience
#OUTPUT - Salary
```

```
#4.DIVIDE THE DATA INTO INPUT and OUTPUT
#INPUT(x) is always 2 dimensional,OUTPUT(y) is always 1 dimensional array
#df.iloc[row slicing,column slicing]
#In df.iloc ,if the column's place has a ':',then array is 2 dimensional
x = df.iloc[0:30,0:1].values
x
#values converts the dataframe into an array
#OUTPUT - Salary
```

```
array([[ 1.1],
       [ 1.3],
       [ 1.5],
       [ 2. ],
       [ 2.2],
       [ 2.9],
       [ 3. ],
       [ 3.2],
       [ 3.2],
       [ 3.7],
       [ 3.9],
```

```
[ ]
       [ 4. ],
       [ 4. ],
       [ 4.1],
       [ 4.5],
       [ 4.9],
       [ 5.1],
       [ 5.3],
       [ 5.9],
       [ 6. ],
       [ 6.8],
       [ 7.1],
       [ 7.9],
       [ 8.2],
       [ 8.7],
       [ 9. ],
       [ 9.5],
       [ 9.6],
       [10.3],
       [10.5]])
```

```
[ ] #In df.iloc ,if the column's place does not has a ':',then array is 1 dimensiona
y = df.iloc[0:30,1].values #df.iloc[:,1].values
#If I want to select all rows or all cols , I can write only :
y
```

```
[ ] array([ 39343., 46205., 37731., 43525., 39891., 56642., 60150.,
          54445., 64445., 57189., 63218., 55794., 56957., 57081.,
          61111., 67938., 66029., 83088., 81363., 93940., 91738.,
          98273., 101302., 113812., 109431., 105582., 116969., 112635.,
          122391., 121872.])
```

```
[ ] #5.TRAIN and TEST VARIABLES
    from sklearn.model_selection import train_test_split
    x_train, x_test, y_train, y_test= train_test_split(x, y, test_size= 1/3, random_state=0)

    # We have 30 observations, so we will take 20 observations for the training set and 10 observations for the test set.
    # We are splitting our dataset so that we can train our model using a training dataset and then test the model using a test dataset.
    #6.NO NEED TO APPLY
```

```
[ ] #7.APPLY REGRESSOR
    from sklearn.linear_model import LinearRegression #Applying regression
    regressor= LinearRegression()
    regressor.fit(x_train, y_train)

    LinearRegression()
```

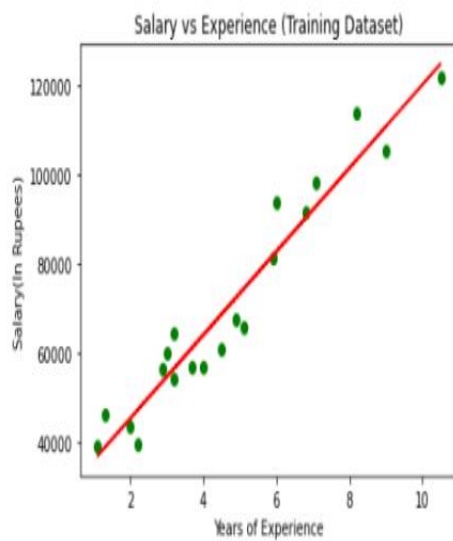
```
[ ] y_pred= regressor.predict(x_test) #Prediction of Test and Training set result
    x_pred= regressor.predict(x_train)
```

```
[ ] import matplotlib.pyplot as plt #Visualizing the Training set results:
```

```
plt.scatter(x_train, y_train, color="green")
plt.plot(x_train, x_pred, color="red")
plt.title("Salary vs Experience (Training Dataset)")
plt.xlabel("Years of Experience")
plt.ylabel("Salary(In Rupees)")
plt.show()
```

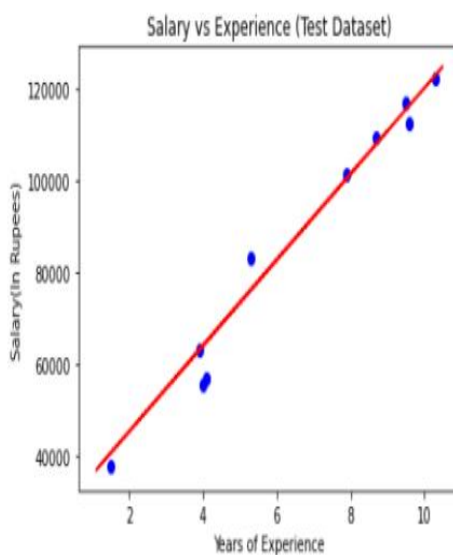
#In the above plot, we can see the real values observations in green dots and predicted values are covered by the red regression line.
#most of the observations are close to the regression line, hence our model is good for the training set.

[]



```
[ ] #Visualizing the Test set results
plt.scatter(x_test, y_test, color="blue")
plt.plot(x_train, x_pred, color="red")
plt.title("Salary vs Experience (Test Dataset)")
plt.xlabel("Years of Experience")
plt.ylabel("Salary(In Rupees)")
plt.show()
```

#In the below plot, there are observations given by the blue color, and prediction is given by the red regression line.
#As we can see, most of the observations are close to the regression line,
#Hence we can say our Simple Linear Regression is a good model and able to make good predictions.



```
[ ] #EDA
```

```
[ ] #EDA
df.head() #displays first 5 rows of dataset
```

	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

```
[ ] df.tail() #displays last 5 rows of dataset
```

	YearsExperience	Salary
25	9.0	105582.0
26	9.5	116969.0
27	9.6	112635.0
28	10.3	122391.0
29	10.5	121872.0

```
[ ] df.describe() #describes dataset
```

	YearsExperience	Salary
count	30.000000	30.000000
mean	5.313333	76003.000000
std	2.837888	27414.429785
min	1.100000	37731.000000
25%	3.200000	56720.750000
50%	4.700000	65237.000000
75%	7.700000	100544.750000
max	10.500000	122391.000000

```
[ ] df.info()
```

```
<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
```

```
[ ] df.shape          #30 rows and 2 column
```

```
(30, 2)
```

```
[ ] df.size           #check the size of dataset
```

```
60
```


Major Project-2: EDA Using 71 Century dataset of Virat Kohli

Major Project 2-EDA.ipynb ☆

File Edit View Insert Runtime Tools Help All changes saved

+ Code + Text

RAM 100% Disk 100% Editing

```
##MAJOR PROJECT##
#EXPLORATORY DATA ANALYSIS USING DATASET#
#EDA ON VIRAT KOHLI CENTURIES SCORED BY HIM IN ALL FORMATS

#1.Take the data and create dataframe
import pandas as pd
df=pd.read_csv('/content/71 Centuries of Virat Kohli.csv')
df
```

	Score	Out/Not Out	Against	Batting Order	Inn.	Strike Rate	Venue	Column1	H/A	Date	Result	Format	Man of the Match	Captain	Unnamed: 14
0	116	Out	Australia	6	2	NaN	Adelaide Oval	Adelaide	Away	24-01-2012	Lost	Test	No	No	NaN
1	103	Out	New Zealand	5	2	NaN	M. Chinnaswamy Stadium	Bangalore	Home	31-08-2012	Won	Test	Yes	No	NaN
2	103	Out	England	5	2	NaN	Vidarbha Cricket Association Stadium	Nagpur	Home	13-12-2012	Drawn	Test	No	No	NaN
3	107	Out	Australia	5	2	NaN	M. A. Chidambaram Stadium	Chennai	Home	22-02-2013	Won	Test	No	No	NaN
4	119	Out	South Africa	4	1	NaN	Wanderers Stadium	Johannesburg	Away	18-12-2013	Drawn	Test	No	No	NaN
...
66	116	Out	Australia	3	1	96.87	Vidarbha Cricket Association Stadium	Nagpur	Home	05-03-2019	Won	ODI	Yes	Yes	NaN
67	123	Out	Australia	3	2	129.47	JSCA International Stadium	Ranchi	Home	08-03-2019	Lost	ODI	No	Yes	NaN
68	120	Out	West Indies	3	1	96.00	Queen's Park Oval	Port of Spain	Away	11-08-2019	Won	ODI	Yes	Yes	NaN
69	114	Not Out	West Indies	3	2	115.15	Queen's Park Oval	Port of Spain	Away	14-08-2019	Won	ODI	Yes	Yes	NaN
70	122	Not Out	Afghanistan	1	1	200.00	Dubai International Cricket Stadium	Dubai	Away	08-09-2022	Won	T20I	Yes	No	NaN

71 rows x 15 columns

```
[ ] df.size #Total no of elements in dataframe
```

1065

```
df.info() # It gives the complete information about the dataframe
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 71 entries, 0 to 70
Data columns (total 15 columns):
#   Column          Non-Null Count  Dtype  
---  -
0   Score            71 non-null    int64  
1   Out/Not Out      71 non-null    object  
2   Against          71 non-null    object  
3   Batting Order    71 non-null    int64  
4   Inn.             71 non-null    int64  
5   Strike Rate      44 non-null    float64 
6   Venue            71 non-null    object  
7   Column1          71 non-null    object  
8   H/A              71 non-null    object  
9   Date             71 non-null    object  
10  Result           71 non-null    object  
11  Format            71 non-null    object  
12  Man of the Match  71 non-null    object  
13  Captain          71 non-null    object  
14  Unnamed: 14      0 non-null     float64 
dtypes: float64(2), int64(3), object(10)
memory usage: 8.4+ KB
```

```
[ ] df.isnull().sum() #To check the null values
```

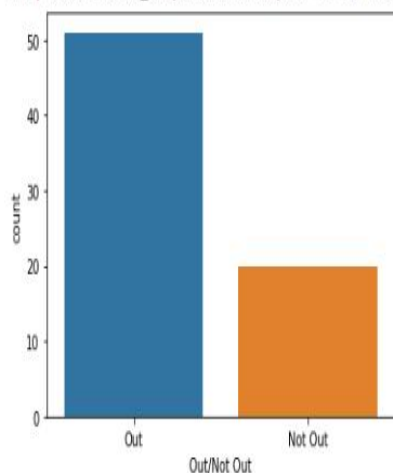
```
Score          0
Out/Not Out    0
Against        0
Batting Order  0
Inn.           0
Strike Rate    27
Venue          0
Column1        0
H/A           0
Date           0
Result         0
Format         0
Man of the Match 0
Captain        0
Unnamed: 14    71
dtype: int64
```

```
[ ] df.nunique()
```

```
Score          48
Out/Not Out     2
Against        10
Batting Order   5
Inn.           4
Strike Rate     44
Venue          45
Column1        42
H/A            2
Date           70
Result         6
Format         3
Man of the Match 2
Captain        2
Unnamed: 14     0
dtype: int64
```

```
[ ] import seaborn as sns #Visualisation-using seaborn
sns.countplot(x='Out/Not Out',data=df) #This Visualisation tells about how many times did virat was remained Notout after hitting century
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f8e01ceabb0>



```
[ ] df['Out/Not Out'].value_counts() #No.of Out/Notout after hitting century
```

```
Out      51
Not Out  20
Name: Out/Not Out, dtype: int64
```

```
[ ] df['Result'].value_counts() #Result when Virat scored a Century
```

```
Won      48
Lost     13
Drawn     7
Lost (D/L)  1
Won (D/L)  1
Tied      1
Name: Result, dtype: int64
```

```
[ ] df['H/A'].value_counts() #Scored 39 centuries in India(Home) and 32 in other country(Away)
```

```
Away     39
Home     32
Name: H/A, dtype: int64
```

```
[ ] df['Format'].value_counts() #No of Centuries scored in terms of Format
```

```
ODI      43
Test     27
T20I      1
Name: Format, dtype: int64
```

```
[ ] df['Captain'].value_counts() #Scored 41 Century as a Captain and 30 as a player
```

```
Yes    41
No     30
Name: Captain, dtype: int64
```

```
[ ] df['Batting Order'].value_counts() #Centuries Scored by Virat On the basis of Batting Order.
```

```
3    36
4    30
5     3
6     1
1     1
Name: Batting Order, dtype: int64
```

```
[ ] df['Against'].value_counts() #No of centuries scored by Virat in terms of Opponent
```

```
Australia    15
Sri Lanka    13
West Indies   11
New Zealand   8
England       8
South Africa  7
Bangladesh    5
Pakistan      2
Zimbabwe      1
```

```
[ ] df['Against'].value_counts()
Afganistan    1
Name: Against, dtype: int64
```

```
[ ] df['year']=pd.to_datetime(df['Date'])
df['year']=df['year'].dt.year
df['Date']=df['Date'].replace('08-09-2022','09-08-2022')
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 71 entries, 0 to 70
Data columns (total 16 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Score            71 non-null    int64
1   Out/Not Out      71 non-null    object
2   Against          71 non-null    object
3   Batting Order    71 non-null    int64
4   Inn.             71 non-null    int64
5   Strike Rate      44 non-null    float64
6   Venue            71 non-null    object
7   Column1          71 non-null    object
8   H/A              71 non-null    object
9   Date             71 non-null    object
10  Result           71 non-null    object
11  Format            71 non-null    object
12  Man of the Match 71 non-null    object
13  Captain          71 non-null    object
```

```
[ ] 14 Unnamed: 14      0 non-null    float64
    15 year           71 non-null    int64
    dtypes: float64(2), int64(4), object(10)
    memory usage: 9.0+ KB
```

```
[ ] df.describe()
```

	Score	Batting Order	Inn.	Strike Rate	Unnamed: 14
count	71.000000	71.000000	71.000000	44.000000	0.0
mean	132.140845	3.521127	1.732394	114.019545	NaN
std	35.911119	0.714326	0.675230	25.257567	NaN
min	100.000000	1.000000	1.000000	84.900000	NaN
25%	107.000000	3.000000	1.000000	96.632500	NaN
50%	119.000000	3.000000	2.000000	108.935000	NaN
75%	139.500000	4.000000	2.000000	120.787500	NaN
max	254.000000	6.000000	4.000000	200.000000	NaN

```
[ ] df.groupby(['Against', 'year'], as_index=False)['Score'].max().sort_values(by='Score', ascending=False).reset_index(drop=True).head() #Kohli Scores Highest (254)
```

	Against	year	Score
0	South Africa	2019	254
1	Sri Lanka	2017	243
2	England	2016	235
3	New Zealand	2016	211
4	Bangladesh	2017	204

```
[ ] df_new = df[['year', 'Against']] #SEPARATE DATA FOR YEAR AND AGAINST COLUMN
df_new
import matplotlib.pyplot as plt
plt.figure(figsize = (30,15))
plt.scatter(df_new['year'], df_new['Against'])

plt.title("Graph Showing Century Scored by Kohli in terms of Opponent and its Year")

plt.show()
```

