The Sparks Foundation Graduate Rotational Internship Program (GRIP) NAME - BARIRA SIDDIQUI DATA SCIENCE AND BUSINESS ANALYTICS INTERN AT THE SPARKS FOUNDATION TASK 1 - PREDICTING USING SUPERVISED MACHINE LEARNING AIM- TO PREDICT THE SCORES OF STUDENTS BASED ON NUMBER OF HOURS STUDIED LINK OF THE DATA- http://bit.ly/w-data **IMPORTING THE LIBRARIES** In [1]: import pandas as pd import numpy as np import seaborn as sbn import matplotlib.pyplot as plt %matplotlib inline **IMPORTING DATA** In [3]: student\_data=pd.read\_csv("http://bit.ly/w-data")#data set stored in variable student\_data student\_data.head(5)#printing the 1st 5 entries of the data set Out[3]: **Hours Scores** 0 2.5 21 1 5.1 47 2 3.2 27 8.5 75 3.5 30 **VISUALIZING THE DATA SET** In [6]: student\_data.shape Out[6]: (25, 2) student\_data.isnull().sum() # checking for null values Out[7]: 0 Hours Scores dtype: int64 student\_data.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 25 entries, 0 to 24 Data columns (total 2 columns): # Column Non-Null Count Dtype float64 0 Hours 25 non-null 1 Scores 25 non-null int64 dtypes: float64(1), int64(1)memory usage: 528.0 bytes In [5]: student\_data.describe() Out[5]: Hours Scores count 25.000000 25.000000 5.012000 51.480000 mean std 2.525094 25.286887 1.100000 17.000000 min 25% 2.700000 30.000000 4.800000 47.000000 **50**% 7.400000 75.000000 9.200000 95.000000 PLOTTING THE SCATTERPLOT In [37]: student\_data.plot.scatter(x='Hours', y='Scores') plt.title('HOURS VS PERCENTAGE') plt.grid() plt.show() HOURS VS PERCENTAGE 90 80 70 £ 60 ĸ 50 40 30 20 Hours PLOTTING DIFFERENT GRAPHS OF THE DATA USING SEABORN In [9]: #Scatterplot sbn.regplot(x=student\_data['Hours'], y=student\_data['Scores']); 100 80 60 40 20 Hours In [10]: #Boxplot sbn.boxplot(data=student\_data[["Hours", "Scores"]]) Out[10]: <AxesSubplot:> 80 60 40 20 0 Hours Scores STORING DATA INTO AN ARRAY In [12]: x=student\_data.iloc[:,:-1].values y=student\_data.iloc[:,1].values In [13]: Out[13]: array([[2.5], [5.1], [3.2], [8.5], [3.5], [1.5], [9.2], [5.5], [8.3], [2.7], [7.7], [5.9], [4.5], [3.3], [1.1], [8.9], [2.5], [1.9], [6.1], [7.4], [2.7], [4.8], [3.8], [6.9], [7.8]]) In [15]: Out[15]: array([21, 47, 27, 75, 30, 20, 88, 60, 81, 25, 85, 62, 41, 42, 17, 95, 30, 24, 67, 69, 30, 54, 35, 76, 86], dtype=int64) SPLITTING THE STUDENT\_DATA FOR TRAINING AND TEST SET In [16]: from sklearn.model\_selection import train\_test\_split x\_train, x\_test, y\_train, y\_test = train\_test\_split(x, y, test\_size=0.25, random\_state=0) #We have allocated 25% of student\_data for testing and 75% of student\_data for training purpose TRAINING THE ALGORITHM **Simple Linear Regression** We will predict the percentage of marks of that a student is expected to score based on number of hours he/she has studied. In [17]: from sklearn.linear\_model import LinearRegression regressor=LinearRegression() regressor.fit(x\_train,y\_train) Out[17]: LinearRegression() **Plotting Regression line** In [19]: line = regressor.coef\_\*x+regressor.intercept\_ student\_data.plot.scatter(x="Hours", y="Scores") plt.scatter(x,y, color= 'yellow'); plt.plot(x, line,color='blue'); plt.grid() plt.show() 80 40 20 Hours The above regression graph clearly shows the positive relation between the scores and hours studied by a student In [21]: y\_pred=regressor.predict(x\_test) print(y\_pred) [16.84472176 33.74557494 75.50062397 26.7864001 60.58810646 39.71058194 20.8213931 ] In [22]: df= pd.DataFrame({'Actual': y\_test, 'Predicted': y\_pred}) Actual Predicted Out[22]: 20 16.844722 0 27 33.745575 69 75.500624 2 30 26.786400 62 60.588106 35 39.710582 24 20.821393 In [23]: sbn.kdeplot(y\_pred, label="Predicted", shade=True); sbn.kdeplot(data=y\_test, label="Actual", shade=True); 0.0200 0.0175 0.0150 0.0125 0.0100 0.0075 0.0050 0.0025 0.0000 PREDICTING THE SCORE OF THE STUDENT STUDIES 9.5 HR/DAY In [34]: hours=9.5 predic=regressor.predict([[hours]]) print("Number of Hours = {}".format(hours)) print("Predicted score = {:.2f}".format(predic[0])) Number of Hours = 9.5Predicted score = 96.38 **Evaluating the Model** In [35]: from sklearn import metrics print('Mean Absolute Error:', metrics.mean\_absolute\_error(y\_test, y\_pred)) Mean Absolute Error: 4.130879918502486