**Iris Dataset** import numpy as np import pandas as pd import matplotlib.pyplot as plt import seaborn as sns df = pd.read csv('iris.csv') df.head() Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Species 3.5 0.2 Iris-setosa 2 4.9 3.0 1.4 0.2 Iris-setosa 3.2 2 4.7 1.3 0.2 Iris-setosa 4.6 3.1 1.5 Iris-setosa 5.0 3.6 1.4 0.2 Iris-setosa **Data Visualization** In [4]: sns.countplot(x=df['Species']) Out[4]: <AxesSubplot:xlabel='Species', ylabel='count'> 50 40 30 count 20 10 Iris-setosa Iris-versicolor Iris-virginica Species sns.pairplot(df,hue='Species') Out[7]: <seaborn.axisgrid.PairGrid at 0x19a3f38e7c0> 150 125 50 25 SepalLengthCm Iris-setosa 3.0 Iris-virginica 2.5 2.0 6 PetalLengthCm 5 2.5 2.0 PetalWidthCm 0.5 100 150 SepalLengthCm PetalWidthCm df.corr() **PetalLengthCm** SepalLengthCm  ${\bf SepalWidthCm}$ **PetalWidthCm** ld 1.000000 0.716676 -0.397729 0.882747 0.899759 SepalLengthCm 0.716676 1.000000 -0.109369 0.871754 0.817954 SepalWidthCm -0.397729 -0.109369 1.000000 -0.420516 -0.356544 **PetalLengthCm** 0.882747 0.871754 -0.420516 1.000000 0.962757 **PetalWidthCm** 0.899759 0.817954 -0.356544 0.962757 1.000000 sns.heatmap(df.corr(),annot=True) <AxesSubplot:> - 1.0 1 0.72 -0.4 0.88 0.9 ld - 0.8 0.6 SepalLengthCm 0.72 1 -0.11 0.87 0.82 0.4 SepalWidthCm -0.4 -0.11 1 -0.42 -0.36 0.2 0.88 0.87 -0.42 1 0.96 PetalLengthCm 0.0 -0.2 0.96 PetalWidthCm 0.9 0.82 -0.36 1 -0.4 p SepalLengthCm **PetalWidthCm** SepalWidthCm PetalLengthCm **Data Cleaning** df.isna().sum() 0 SepalLengthCm 0 SepalWidthCm 0 PetalLengthCm 0 0 Species dtype: int64 no null values Id in the table is not of any use so we will drop it In [14]: df.drop(['Id'],axis=1,inplace=True) Now we need to convert the string values to neumeric by getting dummy or using label Encoder from sklearn.preprocessing import LabelEncoder le = LabelEncoder() le.fit(df['Species']) df['Species'] = le.transform(df['Species']) df.head() SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Species 0 5.1 3.5 0.2 0 1.4 1 4.9 3.0 0.2 0 1.4 2 4.7 3.2 1.3 0.2 0 3 3.1 0.2 0 4.6 1.5 4 5.0 0.2 0 3.6 1.4 Now we have a clean data which can be used to build a model **Model Building** x = df.drop(['Species'],axis=1) y = df['Species'] from sklearn.model\_selection import train\_test\_split x\_train,x\_test,y\_train,y\_test = train\_test\_split(x,y,train\_size=0.7,random\_state=101) 1) Logistic Regression from sklearn.linear\_model import LogisticRegression from sklearn.metrics import accuracy\_score,confusion\_matrix lr = LogisticRegression() lr.fit(x\_train,y\_train) lr\_predict = lr.predict(x\_test) sns.heatmap(confusion\_matrix(y\_test,lr\_predict),annot=True) Out[23]: <AxesSubplot:> - 17.5 0 0 - 15.0 - 12.5 - 10.0 19 - 7.5 - 5.0 0 0 - 2.5 In [24]: accuracy\_score(y\_test,lr\_predict) Out[24]: 0.977777777777777 2) k-Nearest Neighbors from sklearn.neighbors import KNeighborsClassifier kn = KNeighborsClassifier(n neighbors=10) kn.fit(x\_train,y\_train) kn\_predict = kn.predict(x\_test)  $\verb|sns.heatmap| (\verb|confusion_matrix| (\verb|y_test|, \verb|kn_predict|)|, \verb|annot=| True||$ Out[26]: <AxesSubplot:> - 20.0 - 17.5 0 0 - 15.0 - 12.5 - 10.0 - 7.5 - 5.0 0 2.5 0.0 accuracy score(y test,kn predict) Out[27]: 1.0 3) Navie Bayes from sklearn.naive\_bayes import GaussianNB nb = GaussianNB() nb.fit(x\_train,y\_train) nb\_predict = nb.predict(x\_test)  $\verb|sns.heatmap| (\verb|confusion_matrix| (\verb|y_test|, \verb|nb_predict|)|, \verb|annot=| True||$ Out[30]: <AxesSubplot:> - 17.5 0 - 15.0 - 12.5 - 10.0 0 19 7.5 - 5.0 0 1 2.5 0.0 ż accuracy\_score(y\_test,nb\_predict) Out[31]: 0.95555555555556 4)Decision Tree from sklearn.tree import DecisionTreeClassifier dt = DecisionTreeClassifier() dt.fit(x\_train,y\_train) dt\_predict = dt.predict(x\_test) dt\_cm = confusion\_matrix(y\_test,dt\_predict) sns.heatmap(dt\_cm,annot=True) Out[32]: <AxesSubplot:> - 17.5 0 0 0 - 15.0 - 12.5 - 10.0 0 1 19 - 7.5 - 5.0 0 - 2.5 0.0 Ó 1 2 accuracy\_score(y\_test,dt\_predict) Out[33]: 0.95555555555556 5)Random Forest In [34]: from sklearn.ensemble import RandomForestClassifierrf = RandomForestClassifier() rf.fit(x\_train,y\_train) rf\_predict = rf.predict(x\_test) rf\_cm = confusion\_matrix(y\_test,rf\_predict) sns.heatmap(rf\_cm,annot=True) Out[34]: <AxesSubplot:> - 17.5 0 0 - 15.0 - 12.5 - 10.0 0 19 - 7.5 - 5.0 0 - 2.5 accuracy\_score(y\_test,rf\_predict) Out[35]: 0.95555555555556 in all the above Model we got higesht accuracy of KNN model of 100%