IRIS FLOWER CLASSIFICATION: (Task-1) In [1]: #importing basic libraries import pandas as pd import matplotlib.pyplot as plt import seaborn as sns from sklearn.model\_selection import train\_test\_split from sklearn.neighbors import KNeighborsClassifier from sklearn.metrics import classification\_report,accuracy\_score data\_set=pd.read\_csv(r"C:\Users\HP\OneDrive\Documents\oasis infobytes\Iris.csv") data\_set.head()  $SepalLengthCm \quad SepalWidthCm \quad PetalLengthCm \quad PetalWidthCm$ **Species** Out[2]: 0 5.1 3.5 1.4 0.2 Iris-setosa 3.0 1 0.2 Iris-setosa 2 4.7 3.2 1.3 0.2 Iris-setosa 4.6 3.1 1.5 0.2 Iris-setosa 5.0 3.6 1.4 4 0.2 Iris-setosa data\_set.tail() Out[3]: SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm **Species** 295 6.7 3.0 5.2 2.3 Iris-virginica 296 6.3 2.5 5.0 1.9 Iris-virginica 297 6.5 3.0 5.2 2.0 Iris-virginica 298 6.2 3.4 5.4 2.3 Iris-virginica 299 5.9 3.0 5.1 1.8 Iris-virginica data\_set["Species"].unique() array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object) checking missing value data\_set.isnull().sum() SepalLengthCm Out[5]: SepalWidthCm 0 PetalLengthCm 0 0 PetalWidthCm Species 0 dtype: int64 data\_set.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 300 entries, 0 to 299 Data columns (total 5 columns): Column Non-Null Count Dtype SepalLengthCm 300 non-null float64 SepalWidthCm 300 non-null float64 1 PetalLengthCm 300 non-null 2 float64 PetalWidthCm 300 non-null float64 object Species 300 non-null dtypes: float64(4), object(1) memory usage: 11.8+ KB In [7]: type(data\_set) #checking the type of dataset pandas.core.frame.DataFrame Out[7]: In [8]: print(data\_set.shape) #In this csv file we have 300 rows and 5 columns . data\_set.describe() In [9]: Out[9]: SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm 300.000000 300.00000 300.000000 300.000000 count 5.843333 3.054000 1.198667 3.758667 mean std 0.826680 0.432869 1.761467 0.761883 4.300000 1.000000 0.100000 2.000000 min **25**% 5.100000 2.800000 1.600000 0.300000 4.350000 **50**% 5.800000 3.000000 1.300000 **75**% 6.400000 3.300000 5.100000 1.800000 6.900000 2.500000 max 7.900000 4.400000 Data visualizataion In [10]: sns.pairplot(data\_set,hue="Species") plt.show() 8 SepalLengthCm 4.5 4.0 SepalWidthCm 2.5 2.0 Species Iris-setosa Iris-versicolor 6 Iris-virginica PetalLengthCm 2.5 2.0 PetalWidthCm 0.5 0.0 2 0 PetalLengthCm PetalWidthCm SepalLengthCm SepalWidthCm In [11]: sns.lmplot(x='PetalLengthCm', y='PetalWidthCm', data=data\_set) <seaborn.axisgrid.FacetGrid at 0x1448d733a90> 2.5 2.0 PetalWidthCm 1.0 0.5 0.0 2 PetalLengthCm In [12]: sns.heatmap(data\_set.corr(), annot=True) C:\Users\HP\AppData\Local\Temp\ipykernel\_22100\3060311531.py:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will defau It to False. Select only valid columns or specify the value of numeric\_only to silence this warning. sns.heatmap(data\_set.corr(),annot=True) <Axes: > - 1.0 -0.11 0.87 0.82 SepalLengthCm -- 0.8 - 0.6 SepalWidthCm --0.11 -0.42 -0.36 0.4 0.2 -0.42 0.87 1 0.96 PetalLengthCm -0.0 -0.2 PetalWidthCm -0.82 -0.36 0.96 -0.4SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm In [13]: plt.title('Ires flower sepalWidthcm and sepalWidthCm') sns.lineplot(x='SepalLengthCm', y='SepalWidthCm', data=data\_set,color="red") <Axes: title={'center': 'Ires flower sepalWidthcm and sepalWidthCm'}, xlabel='SepalLengthCm', ylabel='SepalWidthCm'> Ires flower sepalWidthcm and sepalWidthCm 3.8 3.6 3.4 SepalWidthCm 3.0 2.8 2.6 2.4 5.5 7.0 7.5 4.5 5.0 6.0 6.5 8.0 SepalLengthCm In [14]: plt.title('Ires flower PetalWidthcm and PetalWidthCm') sns.lineplot(x='PetalLengthCm', y='PetalWidthCm', data=data\_set,color="green") <Axes: title={'center': 'Ires flower PetalWidthcm and PetalWidthCm'}, xlabel='PetalLengthCm', ylabel='PetalWidthCm'> Ires flower PetalWidthcm and PetalWidthCm 2.5 2.0 PetalWidthCm 1.0 0.5 0.0 PetalLengthCm

Modifiying the data In [15]: X=data\_set.drop("Species", axis=1) Out[15]: SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm 5.1 3.5 0.2 4.9 3.0 1.4 0.2 2 4.7 3.2 1.3 0.2 4.6 3.1 1.5 0.2 5.0 3.6 1.4 0.2 2.3 295 6.7 3.0 5.2 296 6.3 2.5 5.0 1.9 6.5 3.0 5.2 2.0 297 3.4 5.4 2.3 298 6.2 3.0 5.1 1.8 299 5.9 300 rows × 4 columns

Finally created model prediction (or) Final Result.

new\_dataset=pd.DataFrame({"SepalLengthCm":2.3, "SepalWidthCm":3.0, "PetalLengthCm":4.2, "PetalWidthCm":7.2}, index=[0])

#Thanking you...

In [21]: **import** pandas **as** pd

Out[21]:

prediction[0]

'Iris-virginica'

prediction=knn.predict(new\_dataset)

In [16]: Y=data\_set["Species"] Iris-setosa Out[16]: Iris-setosa Iris-setosa 2 Iris-setosa 3 Iris-setosa 295 Iris-virginica 296 Iris-virginica 297 Iris-virginica Iris-virginica Iris-virginica Name: Species, Length: 300, dtype: object

Training the model x\_train, x\_test, y\_train, y\_test=train\_test\_split(X, Y, test\_size=0.3, random\_state=42) knn=KNeighborsClassifier(n\_neighbors=3) #KNeihboursClassification Algorithm. In [18]: knn.fit(x\_train, y\_train) Out[18]: ▼ KNeighborsClassifier KNeighborsClassifier(n\_neighbors=3) In [19]: y\_predict=knn.predict(x\_test) In [20]: print("Accuracy:" ,accuracy\_score(y\_test,y\_predict)) Accuracy: 0.9555555555556

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