In [1]: **from** sklearn **import** tree from pandas import read csv from sklearn.tree import DecisionTreeClassifier import numpy as np import matplotlib.pyplot as plt import seaborn as sns df=read csv("Travel.csv") In [2]: df Age EmploymentType GraduateOrNot AnnualIncome FamilyMembers ChronicDiseases FrequentFlyer EverTravelledAbroad TravelInsura Out[2]: Government 0 31 400000 6 Yes 1 No No Sector Private Sector/Self 31 Yes 1250000 No No **Employed** Private Sector/Self 34 4 2 Yes 500000 1 No No **Employed** Private Sector/Self 700000 3 28 Yes 3 No No **Employed** Private Sector/Self 4 28 700000 8 Yes Yes No **Employed** Private Sector/Self 1982 33 Yes 1500000 0 Yes Yes **Employed** Private Sector/Self 1983 28 Yes 1750000 No Yes **Employed** Private Sector/Self 1984 28 6 Yes 1150000 1 No No **Employed** Private Sector/Self 1985 34 Yes 1000000 0 Yes Yes **Employed** Private Sector/Self 1986 Yes 500000 4 0 No No **Employed** 1987 rows × 9 columns In [3]: features=['Age', 'AnnualIncome', 'FamilyMembers','ChronicDiseases'] X=df[features] y=df['TravelInsurance'] print(X) print(y) AnnualIncome FamilyMembers ChronicDiseases 0 31 400000 6 1 7 0 1 31 1250000 1 2 34 500000 4 3 28 700000 3 1 1 4 28 700000 8 1982 33 1500000 4 0 1983 28 1750000 5 1 1984 28 1150000 6 1 0 1985 34 1000000 6 1986 34 500000 0 [1987 rows x 4 columns] 0 0 1 0 2 1 3 0 4 0 1982 1 1983 0 1984 1985 1 1986 0 Name: TravelInsurance, Length: 1987, dtype: int64 inputs=df.drop('TravelInsurance',axis='columns') In [4]: target=df['TravelInsurance'] from sklearn.preprocessing import LabelEncoder In [5]: le\_EmploymentType=LabelEncoder() In [6]: le GraduateOrNot=LabelEncoder() le FrequentFlyer=LabelEncoder() le EverTravelledAbroad=LabelEncoder() inputs['EmploymentType\_n']=le\_EmploymentType.fit\_transform(inputs['EmploymentType']) In [7]: inputs['GraduateOrNot\_n']=le\_GraduateOrNot.fit\_transform(inputs['GraduateOrNot']) inputs['FrequentFlyer\_n']=le\_FrequentFlyer.fit\_transform(inputs['FrequentFlyer']) inputs['EverTravelledAbroad\_n']=le\_EverTravelledAbroad.fit\_transform(inputs['EverTravelledAbroad']) inputs.head() Out[7]: Age EmploymentType GraduateOrNot AnnualIncome FamilyMembers ChronicDiseases FrequentFlyer EverTravelledAbroad EmploymentTy Government 0 31 400000 6 No Sector Private Sector/Self 31 1250000 7 0 No No **Employed** Private Sector/Self 2 34 500000 4 1 No No Yes **Employed** Private Sector/Self 3 28 700000 3 1 No No **Employed** Private Sector/Self 4 28 Yes 700000 8 1 Yes No **Employed** inputs\_n=inputs.drop(['EmploymentType','GraduateOrNot','FrequentFlyer','EverTravelledAbroad'],axis='columns') In [8]: inputs n Out[8]: Age AnnualIncome FamilyMembers ChronicDiseases EmploymentType\_n GraduateOrNot\_n FrequentFlyer\_n EverTravelledAbroad\_n 0 0 31 400000 6 1 0 1 0 1 31 1250000 7 0 0 0 2 34 500000 4 1 1 0 0 3 28 700000 3 0 0 0 4 28 700000 8 1 1 1 1 1982 33 1500000 4 0 1 1 1 1 1983 28 1750000 0 0 1984 28 1150000 6 1 1 1 0 1000000 1985 34 1 0 0 1986 34 500000 4 0 1 1987 rows × 8 columns # Defining the Naive Bayes In [9]: In [10]: from sklearn.naive\_bayes import MultinomialNB mnb=MultinomialNB(fit\_prior=False) import sklearn.model selection from sklearn.model\_selection import train\_test\_split X\_train,X\_test,y\_train,y\_test=train\_test\_split(X,y,test\_size=0.3,random\_state=1) mnb.fit(X\_train,y\_train) MultinomialNB(fit\_prior=False) Out[11]: In [12]: y\_pred=mnb.predict(X\_test) print(y\_pred)  $[0\;0\;0\;0\;1\;1\;0\;0\;0\;1\;0\;1\;0\;0\;0\;1\;0\;0\;0\;1\;0\;1\;0\;1\;1\;0\;1\;1\;1\;0\;0\;0\;0\;0\;0\;0\;1$  $1 \;\; 1 \;\; 0 \;\; 1 \;\; 0 \;\; 1 \;\; 0 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 1 \;\; 1 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 1 \;\; 0 \;\;$  $1 \;\; 1 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 1 \;\; 1 \;\; 0 \;\; 1 \;\; 1 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 1 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 1 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 1 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 1 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 1 \;\; 1 \;\; 0 \;\;$ 0 1 0 1 1] # Confusion Matrix In [13]: In [14]: from sklearn.metrics import confusion\_matrix confusion\_mat =confusion\_matrix(y\_test,y\_pred) sns.heatmap(confusion\_mat,annot=True) plt.xlabel("Predicted Label") plt.ylabel("True label") plt.title("Confusion matrix") plt.show() from sklearn.metrics import confusion matrix cm2=np.array(confusion\_matrix(y\_test,y\_pred)) Confusion matrix - 240 220 2.6e+02 1.3e+02 200 Frue label 180 - 160 140 1.3e+02 80 120 100 i Predicted Label array([[256, 127], Out[14]: [ 80, 134]], dtype=int64) from sklearn.metrics import accuracy score In [15]: accuracy\_score(y\_test,y\_pred) 0.6532663316582915 Out[15]: In [16]: from sklearn.naive bayes import GaussianNB gnb=GaussianNB() gnb.fit(X\_train,y\_train) y\_pred =gnb.predict(X\_test) print(y pred) 0 0 0 0 1] In [17]: **from** sklearn **import** metrics print(accuracy\_score(y\_test,y\_pred)) 0.7738693467336684 In [18]: from sklearn.metrics import confusion matrix cm2=np.array(confusion\_matrix(y\_test,y\_pred)) array([[346, 37], Out[18]: [ 98, 116]], dtype=int64)