1)A)

Output of the given code (Generated Table)

DecisionTreeClassifier	
Dataset 5% 10% 15% 20% 25% 30% 35% 40% 45%	50%
balance-scale 70.10% 72.47% 71.20% 75.69% 73.77% 75.67% 77.74% 75.99% 78.09% 7	82.91% 76.98% 99.20%
BernoulliNB with priors	
Dataset 5% 10% 15% 20% 25% 30% 35% 40% 45%	50%
balance-scale 46.08% 46.08% 46.08% 46.08% 46.08% 46.08% 46.08% 46.08% 46.08% 46.08%	81.28% 46.08% 92.23%

- B) 3 and 5
- C) (1) BNB performs better with priors as default (class_prior = None).
- 2) When Random State is left default:-

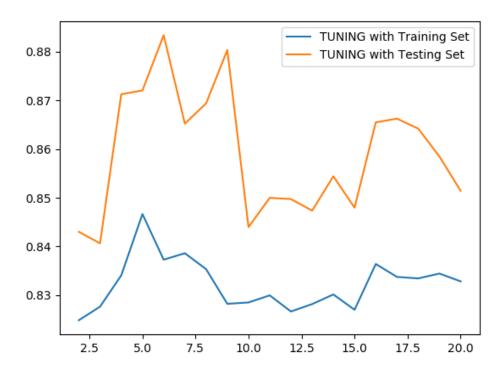
A)

Model accuracy (Testing Set) = 0.8426966292134831 Model accuracy (Training Set) = 0.8564516129032258 ROC_AUC SCORE (Testing Set) = 0.8215654395540587 ROC_AUC SCORE (Training Set) = 0.829507269521243

B)

HYPERPARAMETER SEARCH TUNING WITH TRAINING SET:
Optimal score = 0.8466501506510756
Grid Best Params = {'min_samples_leaf': 5}

HYPERPARAMETER SEARCH TUNING WITH TESTING SET:
Optimal score = 0.8833682446585673
Grid Best Params = {'min_samples_leaf': 6}



D)
PROBABILITY OF A WOMAN FIRST CLASS PASSENGER TO
SURVIVE IS 38.57338017174081

3) When Random_State = 0:-

A)
Model accuracy (Testing Set) = 0.8277153558052435
Model accuracy (Training Set) = 0.864516129032258
ROC_AUC SCORE (Testing Set) = 0.8077601410934745
ROC_AUC SCORE (Training Set) = 0.8380650207665443

B)

HYPERPARAMETER SEARCH TUNING WITH TRAINING SET:

Optimal score = 0.8554181028005216

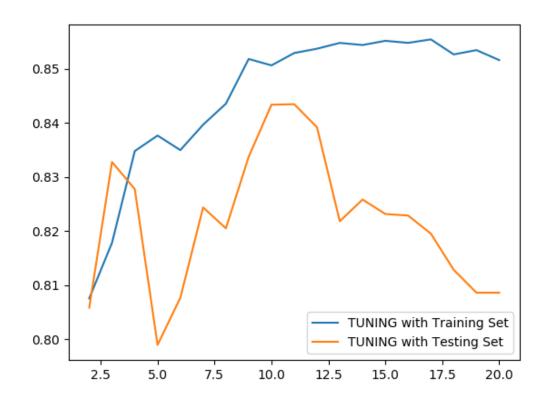
Grid Best Params = {'min_samples_leaf': 17}

HYPERPARAMETER SEARCH TUNING WITH TESTING SET:

Optimal score = 0.8434388528138527

Grid Best Params = {'min_samples_leaf': 11}





D)
PROBABILITY OF A WOMAN FIRST CLASS PASSENGER TO
SURVIVE IS 33.389355742296914

CODE:-

```
from sklearn import tree
from sklearn.model selection import train test split
from sklearn.metrics import roc curve, auc, accuracy score, roc auc score
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import GridSearchCV
"""READING DATASET AND CREATING MODEL"""
df = pd.read csv("titanic.csv")
inputs = df.drop("Survived", axis='columns')
target = df.Survived
normalised_inputs = (inputs - inputs.min())/(inputs.max() - inputs.min())
x train,x test,y train,y test =
train_test_split(inputs, target, test_size=0.3, random_state=0)
treeModel = tree.DecisionTreeClassifier(random state=0)
treeModel.fit(x train,y train)
#MODEL ACCURACY FOR TESTING AND TRAINING SETS
print("Model accuracy (Testing Set) = ",treeModel.score(x_test,y_test))
print("Model accuracy (Training Set) = ",treeModel.score(x_train,y_train))
#ROC AUC SCORE FOR TESTING AND TRAINING SETS
tree_roc_auc_test = roc_auc_score(y_test, treeModel.predict(x_test))
tree_roc_auc_train = roc_auc_score(y_train, treeModel.predict(x_train))
print("ROC_AUC SCORE (Testing Set) = ",tree_roc_auc_test)
print("ROC AUC SCORE (Training Set) = ", tree roc auc train)
#GRID SEARCH ALGO FOR TUNING HYPERPARAMETERS AND SEARCHING OPTIMAL MIN SAMPLES LEAF IN
TESTING AND TRAINING SETS
param grid = dict(min samples leaf=range(2,21))
gridTraining = GridSearchCV(treeModel,param grid,scoring = 'roc auc')
gridTraining.fit(x_train,y_train)
gridTesting = GridSearchCV(treeModel,param_grid,scoring = 'roc_auc')
gridTesting.fit(x_test,y_test)
print("HYPERPARAMETER SEARCH TUNING WITH TRAINING SET :")
print("Optimal score = ", gridTraining.best score )
print("Grid Best Params = ", gridTraining.best params )
print ("HYPERPARAMETER SEARCH TUNING WITH TESTING SET :")
print("Optimal score = ",gridTesting.best_score_)
print("Grid Best Params = ", gridTesting.best params )
plt.plot(range(2,21),gridTraining.cv results ['mean test score'],label='TUNING with
plt.plot(range(2,21),gridTesting.cv_results_['mean_test_score'],label='TUNING with Testing
Set')
plt.legend()
plt.show()
""" PREDICTING POSTERIOR PROBABILITY """
inputSet_for_prediction = []
for i in range(len(inputs['Pclass'])):
    if inputs['Pclass'][i] == 1 and inputs['Sex'][i] == 1:
inputSet for prediction.append([inputs['Pclass'][i],inputs['Sex'][i],inputs['Age'][i],input
ts['Siblings Spouses Aboard'][i],inputs['Parents Children Aboard'][i]])
probs = treeModel.predict_proba(inputSet_for_prediction)[:,1]
print("PROBABILITY OF A WOMAN FIRST CLASS PASSENGER TO SURVIVE IS
", (sum (probs) /len (probs)) *100)
```