CS156 (Introduction to AI), Spring 2022

Final term project

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Project description/introduction text (the background information)

Add the description of this project. Describe the background of the problem your project is trying to solve. Describe the ML problem you are solving below.

Machine learning algorithm selected for this project

Decision Tree & Random Forest

Dataset source

https://www.kaggle.com/competitions/titanic/overview

References and sources

- IF/ELSE All Values = https://stackoverflow.com/questions/44991438/lambda-including-if-elif-else
- Map Values = https://pandas.pydata.org/docs/reference/api/pandas.Series.map.html
- PLT & SNS = https://machinelearningknowledge.ai/seaborn-heatmap-using-sns-heatmap-with-examples-for-beginners/
- Fill In Null Values = https://pandas.pydata.org/docs/reference/api/pandas.DataFrame.fillna.html
- Decision Tree = https://scikit-learn.org/stable/modules/tree.html#:~:text=Decision%20Trees %20(DTs)%20are%20a,as%20a%20piecewise%20constant%20approximation.
- $\label{eq:Graphviz} \textbf{Graphviz} = \text{https://towardsdatascience.com/visualizing-decision-trees-with-python-scikit-learn-graphviz-matplotlib-1c50b4aa68dc}$
- Check call & PImage = https://stackoverflow.com/questions/5316206/converting-dot-to-png-in-python
- Google Colab Output File = https://cyublog.com/articles/python-en/colab-pandas-three-ways-to-save-dataframe-data/
- Extra = https://towardsdatascience.com/visualizing-decision-trees-in-jupyter-notebook-with-python-and-graphviz- 78703230a7b1

Solution

```
Load libraries and set random number generator seed
import numpy as np
import pandas as pd
from sklearn import tree
from sklearn.ensemble import RandomForestClassifier
from sklearn model selection import KFold
from subprocess import check call
from IPython.display import Image as PImage
from google.colab import files
import graphviz
import io
import seaborn as sns
import matplotlib.pyplot as plt
np.random.seed(42)
Code the solution
###Making Titanic Dataset
# Loading the data
uploaded = files.upload()
train = pd.read csv(io.BytesIO(uploaded['train.csv']))
uploaded2 = files.upload()
test = pd.read csv(io.BytesIO(uploaded2['test.csv']))
<IPython.core.display.HTML object>
Saving train.csv to train (2).csv
<IPython.core.display.HTML object>
Saving test.csv to test (2).csv
train.head()
   PassengerId
                Survived Pclass \
0
             1
                       0
                                3
1
             2
                       1
                                1
2
             3
                       1
                                3
3
             4
                       1
                                1
             5
4
                                3
                                                 Name
                                                           Sex
                                                                 Age
SibSp \
                              Braund, Mr. Owen Harris
                                                          male 22.0
1
  Cumings, Mrs. John Bradley (Florence Briggs Th... female 38.0
1
2
                               Heikkinen, Miss. Laina female 26.0
```

```
0
3
        Futrelle, Mrs. Jacques Heath (Lily May Peel) female 35.0
1
4
                             Allen, Mr. William Henry
                                                          male 35.0
0
   Parch
                     Ticket
                                Fare Cabin Embarked
0
       0
                 A/5 21171
                              7.2500
                                       NaN
                                                   S
                                                   C
                                       C85
1
       0
                  PC 17599
                             71.2833
2
                                                   S
       0
         STON/02. 3101282
                              7.9250
                                       NaN
                                                   S
3
       0
                     113803
                             53.1000
                                      C123
4
                                                   S
       0
                     373450
                              8.0500
                                       NaN
test.head() #test does not know who survives
   PassengerId Pclass
                                                                   Name
Sex \
           892
0
                      3
                                                      Kelly, Mr. James
male
                                     Wilkes, Mrs. James (Ellen Needs)
           893
                      3
female
                      2
                                             Myles, Mr. Thomas Francis
           894
male
           895
                      3
                                                      Wirz, Mr. Albert
male
           896
                      3
                         Hirvonen, Mrs. Alexander (Helga E Lindqvist)
female
    Age
         SibSp
               Parch
                         Ticket
                                    Fare Cabin Embarked
  34.5
             0
                     0
                         330911
                                  7.8292
                                            NaN
                                                       0
                                                       S
  47.0
                         363272
1
             1
                                  7.0000
                                            NaN
                     0
                                                       Q
2
  62.0
             0
                     0
                         240276
                                  9.6875
                                            NaN
3
  27.0
             0
                         315154
                                  8.6625
                                                       S
                    0
                                            NaN
                                                       S
4 22.0
             1
                    1
                        3101298
                                 12.2875
                                            NaN
#Testing Use
train2 = train.copy()
test2 = test.copy()
###Add Columns, Fix Nulls, Remove Unnecessary Variables, & Plot Seaborne
#Testing Use
train = train2.copy()
test = test2.copy()
#Adds a column where...
#value is 1 if there is a cabin, 0 otherwise
train['Has Cabin'] = train["Cabin"].apply(lambda x: 0 if type(x) ==
float else 1)
#it combines SibSp and Parch to find family size
```

```
train['Family_Size'] = train['SibSp'] + train['Parch'] + 1
train
```

0 1 2 3 4	Passen	gerId Su 1 2 3 4 5	orvived 0 1 1 1 0	Pclass 3 1 3 1 3	\						
886 887 888 889 890		887 888 889 890 891	0 1 0 1	2 1 3 1 3							
Circ	\						Name	e Sex	Age		
SibS 0	p \			Braur	nd, Mr	. Ower	n Harris	s male	22.0		
1	Cumings, Mrs. John Bradley (Florence Briggs Th								38.0		
1				Heil	kkinen	, Miss	s. Laina	a female	26.0		
0	Futrelle, Mrs. Jacques Heath (Lily May Peel)								35.0		
1 4 0				Allen	, Mr. N	Willia	am Henry	/ male	35.0		
886				Mon	ntvila	, Rev	. Juozas	s male	27.0		
0 887	Graham, Miss. Margaret Edith female 19.0										
0 888		' female	NaN								
1 889				Bel	nr, Mr	. Kar	l Howell	l male	26.0		
0 890 0				[Dooley	, Mr.	Patrick	c male	32.0		
	Parch		Ticket	Fa	re Cab	in Emb	parked	Has_Cabin	1		
0	ly_Size 0	A	A/5 21171	7.250	90 Na	aN	S	e)		
2	0		71.283	33 C	85	С	1				
2	0	STON/02.	3101282	7.925	50 Na	aN	S	6)		
1 3	0		113803	53.100	00 C1	23	S	1			

```
2
4
                       373450
                                 8.0500
                                           NaN
                                                       S
                                                                  0
         0
1
. .
                                    . . .
                                           . . .
                                                     . . .
        . . .
                           . . .
                                                                 . . .
886
         0
                       211536
                                13.0000
                                           NaN
                                                       S
                                                                  0
1
887
         0
                       112053
                                30.0000
                                           B42
                                                       S
                                                                  1
1
                   W./C. 6607 23.4500
888
         2
                                           NaN
                                                       S
                                                                  0
4
889
         0
                       111369
                                30.0000
                                          C148
                                                       C
                                                                  1
890
                       370376
                                                       0
                                                                  0
         0
                                 7.7500
                                           NaN
1
[891 rows x 14 columns]
#Replaces Null Values in Embarked, Fare, & Age With the Average Value
train['Embarked'] = train['Embarked'].fillna('S')
train['Fare'] = train['Fare'].fillna(train['Fare'].median())
train['Age'] = train['Age'].fillna(train['Age'].median())
train
     PassengerId
                   Survived Pclass \
0
                                   3
                1
1
                2
                           1
                                   1
2
                3
                                   3
                           1
3
                4
                           1
                                   1
4
                5
                           0
                                   3
. .
                                 . . .
              . . .
                         . . .
                                   2
886
              887
                           0
887
              888
                           1
                                   1
                                   3
888
              889
                           0
                                   1
              890
                           1
889
890
              891
                           0
                                   3
                                                      Name
                                                               Sex
                                                                      Age
SibSp \
                                 Braund, Mr. Owen Harris
                                                              male 22.0
1
1
     Cumings, Mrs. John Bradley (Florence Briggs Th...
                                                            female 38.0
1
2
                                  Heikkinen, Miss. Laina
                                                            female
                                                                    26.0
0
3
           Futrelle, Mrs. Jacques Heath (Lily May Peel)
                                                            female 35.0
1
4
                                Allen, Mr. William Henry
                                                                    35.0
                                                              male
0
```

```
Montvila, Rev. Juozas
886
                                                            male 27.0
0
887
                           Graham, Miss. Margaret Edith
                                                          female
                                                                  19.0
0
888
              Johnston, Miss. Catherine Helen "Carrie"
                                                          female 28.0
1
889
                                  Behr, Mr. Karl Howell
                                                            male 26.0
0
890
                                    Dooley, Mr. Patrick
                                                            male 32.0
0
     Parch
                      Ticket
                                  Fare Cabin Embarked Has Cabin
Family Size
                   A/5 21171
                                7.2500
                                         NaN
                                                     S
                                                                 0
2
                    PC 17599
                                                     C
1
         0
                               71.2833
                                         C85
                                                                 1
2
2
            STON/02. 3101282
                                                     S
         0
                               7.9250
                                         NaN
                                                                 0
1
3
         0
                       113803
                               53.1000
                                        C123
                                                     S
                                                                 1
2
4
         0
                       373450
                                8.0500
                                         NaN
                                                     S
                                                                 0
1
. .
       . . .
                          . . .
                                   . . .
                                          . . .
                                                   . . .
886
         0
                       211536
                               13.0000
                                         NaN
                                                     S
                                                                 0
1
887
                                                     S
                                                                 1
         0
                       112053
                               30.0000
                                         B42
1
         2
                  W./C. 6607
                                                     S
                                                                 0
888
                               23.4500
                                         NaN
4
889
                       111369
                               30.0000
                                                     C
                                                                 1
         0
                                        C148
1
890
         0
                       370376
                                7.7500
                                                     0
                                                                 0
                                         NaN
1
[891 rows x 14 columns]
# Map Sex
if (train['Sex'][0] != 0) and (train['Sex'][0] != 1):
  train['Sex'] = train['Sex'].map( {'female': 0, 'male':
1} ).astype(int)
# Map Embarked
if (train['Embarked'][0] != 0) and (train['Embarked'][0] != 1) and
(train['Embarked'][0] != 2):
  train['Embarked'] = train['Embarked'].map( {'S': 0, 'C': 1, 'Q':
2} ).astype(int)
```

```
# Map Fare
fare min = train['Fare'].min()
fare q1 = train['Fare'].quantile(.25)
fare median = train['Fare'].median()
fare q3 = train['Fare'].quantile(.75)
fare max = train['Fare'].max()
fare_min, fare_q1, fare_median, fare q3, fare max
train.loc[(train['Fare'] >= fare min) \& (train['Fare'] <= fare q1),
'Fare'] = 0
train.loc[(train['Fare'] > fare q1) & (train['Fare'] < fare median),</pre>
'Fare'] = 1
train.loc[(train['Fare'] > fare median) & (train['Fare'] < fare q3),</pre>
'Fare'] = 2
train.loc[(train['Fare'] >= fare q3) & (train['Fare'] <= fare max),</pre>
Fare' = 3
# Map Age
age min = train['Age'].min()
age_q1 = train['Age'].quantile(.25)
age median = train['Age'].median()
age_q3 = train['Age'].quantile(.75)
age max = train['Age'].max()
age min, age q1, age median, age q3, age max
train.loc[(train['Age'] >= age_min) & (train['Age'] <= age_q1), 'Age']</pre>
train.loc[(train['Age'] > age q1) & (train['Age'] < age median),</pre>
'Age'] = 1
train.loc[(train['Age'] > age median) & (train['Age'] < age q3),
'Age'] = 2
train.loc[(train['Age'] >= age q3) & (train['Age'] <= age max), 'Age']</pre>
= 3
train
     PassengerId Survived
                            Pclass \
0
                          0
                                   3
               1
1
               2
                          1
                                   1
2
                                   3
               3
                          1
3
               4
                          1
                                   1
               5
4
                          0
                                   3
886
             887
                          0
                                   2
                                   1
887
             888
                          1
                                   3
888
             889
                          0
889
             890
                          1
                                   1
                                   3
890
             891
                          0
                                                     Name Sex
                                                                 Age
```

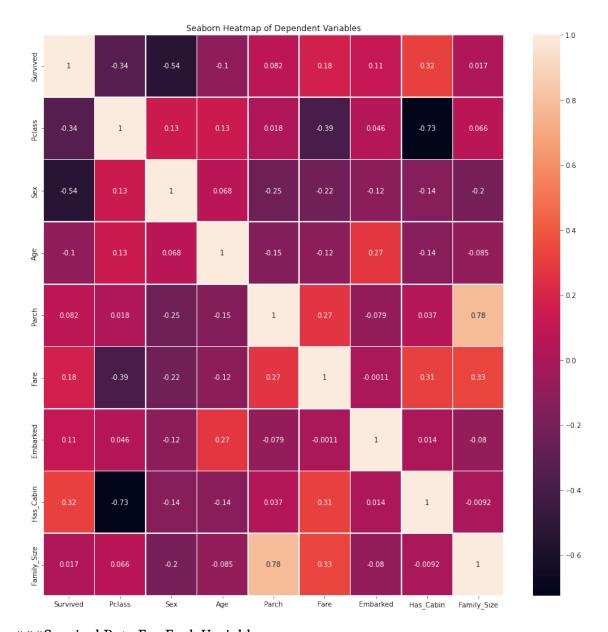
SibSp \

0		ris 1	0.0							
1 1 2	Cuming	s, Mrs. John Brad	ley (Fl	orence.	Briggs Th	0	3.0			
			Hei	kkinen	, Miss. La	ina 0	1.0			
0 3	Futrelle, Mrs. Jacques Heath (Lily May Peel) 0									
1 4		nry 1	3.0							
0 										
886			Мс	ntvila	, Rev. Juo	zas 1	1.0			
0 887	Graham, Miss. Margaret Edith 0 0.0									
0 888		Johnston, Miss	. Cathe	erine H	elen "Carr	ie" 0	28.0			
1 889	Behr, Mr. Karl Howell 1 1.0									
0 890 0				Dooley	, Mr. Patr	ick 1	2.0			
U	Parch	Ti ckot	Fara	Cabin	Embarkad	Was Cabi	n			
	ly_Size	Ticket			Embarked	_				
0 2	0	A/5 21171	0.0	NaN	0		0			
1 2	0	PC 17599	3.0	C85	1		1			
2	0	STON/02. 3101282	1.0	NaN	0		0			
3	0	113803	3.0	C123	0		1			
4 1	0	373450	1.0	NaN	0		0			
886	0	211536	1.0	NaN	0		0			
1 887	0	112053	2.0	B42	0		1			
1 888	2	W./C. 6607	2.0	NaN	0		0			
4 889	0	111369	2.0	C148	1		1			
1 890 1	0	370376	0.0	NaN	2		0			

[891 rows x 14 columns]

```
#Drop columns that are not needed in calculation
try:
  PassengerId = test['PassengerId'] #Save ID for future use
  dropColumns = ['PassengerId', 'Name', 'Ticket', 'Cabin', 'SibSp']
  train = train.drop(dropColumns, axis = 1)
  train
except:
  print("Already Dropped")
#Do the same for test
#Adds a column where...
#value is 1 if there is a cabin, 0 otherwise
test['Has Cabin'] = test["Cabin"].apply(lambda x: 0 if type(x) ==
float else 1)
#it combines SibSp and Parch to find family size
test['Family Size'] = test['SibSp'] + test['Parch'] + 1
#Replaces Null Values in Embarked, Fare, & Age With the Average Value
test['Embarked'] = test['Embarked'].fillna('S')
test['Fare'] = test['Fare'].fillna(test['Fare'].median())
test['Age'] = test['Age'].fillna(test['Age'].median())
test
#Replaces Null Values in Embarked, Fare, & Age With the Average Value
test['Embarked'] = test['Embarked'].fillna('S')
test['Fare'] = test['Fare'].fillna(test['Fare'].median())
test['Age'] = test['Age'].fillna(test['Age'].median())
test
# Map Sex
if (test['Sex'][0] != 0) and (test['Sex'][0] != 1):
test['Sex'] = test['Sex'].map( {'female': 0, 'male': 1} ).astype(int)
# Map Embarked
if (\text{test}['\text{Embarked}'][0] != 0) and (\text{test}['\text{Embarked}'][0] != 1) and
(test['Embarked'][0] != 2):
test['Embarked'] = test['Embarked'].map( {'S': 0, 'C': 1, 'Q':
2} ).astype(int)
# Map Fare
fare_min = test['Fare'].min()
fare g1 = test['Fare'].guantile(.25)
fare median = test['Fare'].median()
fare q3 = test['Fare'].quantile(.75)
fare max = test['Fare'].max()
fare_min, fare_q1, fare_median, fare_q3, fare_max
test.loc[(test['Fare'] >= fare min) & (test['Fare'] <= fare q1),
'Fare'1 = 0
test.loc[(test['Fare'] > fare q1) & (test['Fare'] < fare median),
'Fare'] = 1
test.loc[(test['Fare'] > fare median) & (test['Fare'] < fare q3),
'Fare'] = 2
```

```
test.loc[(test['Fare'] >= fare g3) & (test['Fare'] <= fare max),</pre>
'Fare'l = 3
# Map Age
age min = test['Age'].min()
age q1 = test['Age'].quantile(.25)
age median = test['Age'].median()
age q3 = test['Age'].quantile(.75)
age max = test['Age'].max()
age min, age q1, age median, age q3, age max
test.loc[(test['Age'] \geq age min) & (test['Age'] \leq age q1), 'Age'] =
test.loc[(test['Age'] > age q1) & (test['Age'] < age median), 'Age'] =
test.loc[(test['Age'] > age median) & (test['Age'] < age q3), 'Age'] =
test.loc[(test['Age'] >= age_q3) & (test['Age'] <= age_max), 'Age'] =</pre>
test
#Drop columns that are not needed in calculation
try:
dropColumns = ['PassengerId', 'Name', 'Ticket', 'Cabin', 'SibSp']
test = test.drop(dropColumns, axis = 1)
 test
except:
print("Already Dropped")
plt.figure(figsize=(15,15))
plt.title('Seaborn Heatmap of Dependent Variables')
sns.heatmap(train.astype(float).corr(),linewidths=0.6, annot=True)
<matplotlib.axes. subplots.AxesSubplot at 0x7f6a01e6ad50>
```



###Survival Rate For Each Variable

```
#mean = survival rate, count = total people, sum = survivors
train[['Pclass', 'Survived']].groupby(['Pclass'],
as_index=False).agg(['mean', 'count', 'sum'])
```

```
Survived
            mean count
                        sum
Pclass
        0.629630
                   216
                         136
1
2
        0.472826
                   184
                         87
3
        0.242363
                   491
                        119
```

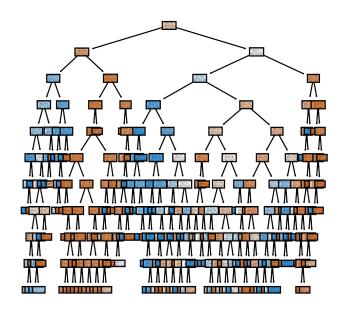
```
#mean = survival rate, count = total people, sum = survivors
train[['Sex', 'Survived']].groupby(['Sex'],
as_index=False).agg(['mean', 'count', 'sum'])
     Survived
         mean count sum
Sex
0
     0.742038
                314
                     233
1
     0.188908
                577
                     109
train[['Age', 'Survived']].groupby(['Age'],
as_index=False).agg(['mean', 'count', 'sum'])
      Survived
          mean count sum
Age
0.0
      0.424242
                 231
                     98
1.0
      0.405660
                     43
                 106
2.0
      0.410256
                 117 48
3.0
      0.400000
                 235
                      94
28.0 0.292079
                 202
                     59
train[['Parch', 'Survived']].groupby(['Parch'],
as_index=False).agg(['mean', 'count', 'sum'])
       Survived
           mean count sum
Parch
0
       0.343658
                  678 233
1
       0.550847
                  118
                        65
2
       0.500000
                   80
                        40
3
       0.600000
                    5
                         3
4
       0.000000
                    4
                         0
5
       0.200000
                    5
                         1
       0.000000
                         0
train[['Embarked', 'Survived']].groupby(['Embarked'],
as index=False).agg(['mean', 'count', 'sum'])
          Survived
              mean count sum
Embarked
          0.339009
                     646
                          219
0
1
          0.553571
                     168
                           93
2
          0.389610
                      77
                           30
train[['Has Cabin', 'Survived']].groupby(['Has_Cabin'],
as_index=False).agg(['mean', 'count', 'sum'])
           Survived
               mean count sum
Has Cabin
```

```
687
0
           0.299854
                            206
1
           0.666667
                      204
                            136
train[['Family Size', 'Survived']].groupby(['Family_Size'],
as index=False).agg(['mean', 'count', 'sum'])
             Survived
                 mean count
                              sum
Family_Size
             0.303538
                         537
                              163
2
             0.552795
                         161
                               89
3
             0.578431
                         102
                               59
4
                          29
             0.724138
                               21
5
             0.200000
                          15
                                3
6
             0.136364
                          22
                                3
7
             0.333333
                          12
                                4
8
             0.000000
                          6
                                0
11
             0.000000
                           7
                                0
###Decision Tree
#Use Decision Tree
accuracyList = list()
for x in range(1, 12):
    accuracyFold = []
    kf = KFold(n splits=10)
    clf = tree.DecisionTreeClassifier(max depth = x)
    for trainFold, testFold in kf.split(train):
        trainData = train.loc[trainFold]
        testData = train.loc[testFold]
        model = clf.fit(X = trainData.drop(['Survived'], axis=1), y =
trainData["Survived"])
        testAccount = model.score(X = testData.drop(['Survived'],
axis=1), y = testData["Survived"])
        accuracyFold.append(testAccount)
    accuracyList.append(sum(accuracyFold)/len(accuracyFold))
df = pd.DataFrame({"Max Depth":range(1,12), "Average Accuracy":
accuracyList})
df = df[["Max Depth", "Average Accuracy"]]
df
    Max Depth Average Accuracy
0
            1
                        0.786729
            2
1
                        0.771099
2
            3
                        0.797990
3
            4
                        0.805855
4
            5
                        0.805843
5
            6
                        0.806979
6
            7
                        0.809201
7
            8
                        0.811473
```

```
9
                       0.800237
8
9
           10
                       0.786804
10
           11
                       0.793496
y train = train['Survived']
x train = train.drop(['Survived'], axis=1).values
x test = test.values
#Checks with max depth should be used
Max depth = 0;
for row in df.index:
  if df['Average Accuracy'][row] == max(accuracyList):
    Max depth = row
# Decision with max depth
decision tree = tree.DecisionTreeClassifier(max depth = Max depth)
decision tree.fit(x train, y train)
# Predicting results for test dataset
prediction = decision tree.predict(x test)
submission = pd.DataFrame({
        "PassengerId": PassengerId,
        "Survived": prediction
    })
survivors = 0
for x in submission['Survived']:
  if (x == 1):
    survivors = survivors + 1
print(str(survivors) + " will survive out of " +
str(submission.index.size))
168 will survive out of 418
#Save to file
submission.to csv('submission.csv', encoding = 'utf-8-sig')
files.download('submission.csv')
<IPython.core.display.Javascript object>
<IPython.core.display.Javascript object>
dot_data = tree.export_graphviz(decision_tree, out_file=None)
graph = graphviz.Source(dot_data)
graph.render("Titanic")
dot data = tree.export graphviz(decision tree, out file=None,
                      feature names = list(train.drop(['Survived'],
axis=1)).
                      class_names = ['Died', 'Survived'],
                     filled=True, rounded=True,
                     special characters=True)
```

```
graph = graphviz.Source(dot data)
#Display Graph as Image
img = graph.save('image.png')
check call(['dot','-Tpng','tree1.dot','-o','image.png'])
PImage(img)
                                                   Sex <= 0.5
gini = 0.473
samples = 891
value = [549, 342]
class = Died
                                                                Has_Cabin <= 0.5
gini = 0.306
samples = 577
value = [468, 109]
class = Died
                                       Pclass <= 2.5
gini = 0.383
samples = 314
value = [81, 233]
class = Survived
                                       Family_Size <= 4.5
gini = 0.5
samples = 144
value = [72, 72]
class = Died
                                                                                   Fare <= 0.5
gini = 0.487
samples = 107
value = [62, 45]
class = Died
                                                                Age <= 2.5
gini = 0.235
samples = 470
value = [406, 64]
class = Died
                                                                                               Age <= 1.5
gini = 0.496
samples = 99
value = [54, 45]
class = Died
                                         = 0.5
                                 Embarked s
                                                Fare <= 1.5
gini = 0.198
samples = 27
value = [24, 3
                                  gini = 0.484
samples = 117
value = [48, 69]
class = Survived
         ini = 0.17
                        = 0.041
                                  (...)
  (...)
         (...)
                   (...)
                         (...)
                                       (...)
                                                (...)
                                                      (...)
                                                             (...)
                                                                    (...)
                                                                          (...)
                                                                                 (...)
                                                                                              (...)
                                                                                                     (...)
totalAccuracyScore = decision tree.score(x train, y train)
totalAccuracyScore
0.8653198653198653
###Random Forest Classifier
#Use Random Forest Tree
accuracyList = list()
for x in range(1, 12):
      accuracyFold = []
      kf = KFold(n splits=10)
      clf = RandomForestClassifier(max depth = x, random state = 0)
      for trainFold, testFold in kf.split(train):
             trainData = train.loc[trainFold]
             testData = train.loc[testFold]
             model = clf.fit(X = trainData.drop(['Survived'], axis=1), y =
trainData["Survived"])
             testAccount = model.score(X = testData.drop(['Survived'],
axis=1), y = testData["Survived"])
             accuracyFold.append(testAccount)
      accuracyList.append(sum(accuracyFold)/len(accuracyFold))
df = pd.DataFrame({"Max Depth":range(1,12), "Average Accuracy":
accuracyList})
df = df[["Max Depth", "Average Accuracy"]]
df
```

```
Max Depth Average Accuracy
0
                       0.741935
            1
            2
1
                       0.777890
2
            3
                       0.778989
3
            4
                       0.800300
4
            5
                       0.805868
5
            6
                       0.806979
6
            7
                       0.811461
7
            8
                       0.808065
8
            9
                       0.803620
9
           10
                       0.809226
10
           11
                       0.814844
y_train = train['Survived']
x train = train.drop(['Survived'], axis=1).values
x test = test.values
#Checks with max depth should be used
Max depth = 0;
for row in df.index:
  if df['Average Accuracy'][row] == max(accuracyList):
    Max depth = row
# Decision with max depth
randomForest = RandomForestClassifier(max depth = Max depth)
randomForest.fit(x train, y train)
# Predicting results for test dataset
prediction = decision tree.predict(x test)
submission = pd.DataFrame({
        "PassengerId": PassengerId,
        "Survived": prediction
    })
survivors = 0
for x in submission['Survived']:
  if (x == 1):
    survivors = survivors + 1
print(str(survivors) + " will survive out of " +
str(submission.index.size))
168 will survive out of 418
#Save to file
submission.to csv('submission2.csv', encoding = 'utf-8-sig')
files.download('submission2.csv')
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```



totalAccuracyScore = randomForest.score(x_train, y_train)
totalAccuracyScore

0.8945005611672279