xo8kodezx

April 10, 2024

```
[10]: import os
      import pandas as pnd
      import numpy as nupy
      import seaborn as sns
      import matplotlib.pyplot as plot
[11]: url1 = 'https://raw.githubusercontent.com/Shourya0712/titanicDT/main/
       ⇔titanic_train.csv'
      url2 = 'https://raw.githubusercontent.com/Shourya0712/titanicDT/main/test.csv'
      train_set = pnd.read_csv(url1)
      test_set = pnd.read_csv(url2)
[12]: train_set.head()
[12]:
         PassengerId
                      Survived Pclass
                   1
                              0
      1
                   2
                              1
                                      1
      2
                   3
                              1
                                      3
                   4
      3
                              1
                                      1
                   5
                              0
      4
                                      3
                                                        Name
                                                                 Sex
                                                                       Age SibSp \
      0
                                    Braund, Mr. Owen Harris
                                                                male
                                                                      22.0
                                                                                 1
         Cumings, Mrs. John Bradley (Florence Briggs Th... female 38.0
      1
                                                                               1
      2
                                     Heikkinen, Miss. Laina
                                                              female
                                                                                 0
      3
              Futrelle, Mrs. Jacques Heath (Lily May Peel)
                                                                      35.0
                                                              female
                                                                                 1
      4
                                   Allen, Mr. William Henry
                                                                                 0
                                                                male 35.0
                                      Fare Cabin Embarked
         Parch
                          Ticket
      0
             0
                       A/5 21171
                                    7.2500
                                             NaN
                                                         S
                                                         С
      1
                        PC 17599
                                   71.2833
                                             C85
             0
                                                         S
      2
                STON/02. 3101282
                                    7.9250
                                             NaN
                                                         S
      3
                          113803
                                   53.1000 C123
      4
             0
                          373450
                                    8.0500
                                                         S
                                             NaN
[14]: train_set.info()
```

<class 'pandas.core.frame.DataFrame'>

Data columns (total 12 columns): # Non-Null Count Column Dtype PassengerId 891 non-null int64 0 1 Survived 891 non-null int64 2 Pclass 891 non-null int64 Name 891 non-null object 4 Sex 891 non-null object 5 Age 714 non-null float64 6 SibSp 891 non-null int64 7 Parch 891 non-null int64 8 Ticket 891 non-null object 9 Fare 891 non-null float64 10 Cabin 204 non-null object 889 non-null 11 Embarked object dtypes: float64(2), int64(5), object(5) memory usage: 83.7+ KB [15]: train_set.isnull().sum() [15]: PassengerId 0 Survived 0 Pclass 0 Name 0 Sex 0 Age 177 SibSp 0 Parch 0 Ticket 0 Fare 0 Cabin 687 Embarked 2 dtype: int64 [16]: train_set['Age'] = train_set['Age'].fillna(train_set['Age'].mean()) [18]: train_set.isnull().sum() [18]: PassengerId 0 Survived 0 Pclass 0 Name 0 Sex 0 Age 0 0 SibSp Parch 0

RangeIndex: 891 entries, 0 to 890

```
Ticket
                       0
      Fare
                       0
      Cabin
                     687
      Embarked
      dtype: int64
[19]: train_set['Embarked'] = train_set['Embarked'].fillna(train_set['Embarked'].
       →mode()[0])
[20]: train_set.isnull().sum()
[20]: PassengerId
                       0
      Survived
                        0
      Pclass
                        0
      Name
                        0
      Sex
                       0
      Age
                        0
      SibSp
                        0
      Parch
      Ticket
                       0
      Fare
                        0
      Cabin
                     687
      Embarked
                       0
      dtype: int64
[21]: embarked_mode = train_set['Embarked'].mode()[0]
      print(embarked_mode)
     S
[22]: train_set['Embarked'] = train_set['Embarked'].fillna(embarked_mode)
[23]: train_set.isnull().sum()
[23]: PassengerId
                       0
      Survived
                        0
      Pclass
                        0
      Name
                       0
      Sex
                       0
      Age
                        0
      SibSp
                        0
      Parch
                        0
      Ticket
                        0
      Fare
                        0
      Cabin
                     687
      Embarked
                       0
      dtype: int64
```

```
[24]: train_set['Cabin_Letter'] = train_set['Cabin'].apply(lambda x: str(x)[0])
[25]: train_set['Cabin_Letter'].value_counts()
[25]: Cabin_Letter
     n
          687
     С
           59
     В
           47
     D
           33
     Ε
           32
     Α
           15
     F
           13
            4
     Т
            1
     Name: count, dtype: int64
[26]: X = train_set[['Pclass', 'Sex', 'Age', 'SibSp', 'Parch', 'Fare', __
      y = train_set['Survived']
[27]: from sklearn.preprocessing import LabelEncoder
     from sklearn.model_selection import train_test_split
     from sklearn.tree import DecisionTreeClassifier
     from sklearn import tree
[28]: le = LabelEncoder()
     train_set['Sex'] = le.fit_transform(train_set['Sex'])
[29]: X = train_set[['Pclass', 'Sex', 'Age', 'SibSp', 'Parch', 'Fare', \( \)
      y = train_set['Survived']
[30]: x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.2,_u
       →random_state=75)
[31]: train_set['Embarked'] = le.fit_transform(train_set['Embarked'])
[32]: train_set['Cabin_Letter'] = le.fit_transform(train_set['Cabin_Letter'])
[33]: train set.head(25)
[33]:
         PassengerId Survived Pclass \
     0
                   1
                            0
                                    3
                   2
                             1
     1
                                    1
     2
                   3
                             1
                                    3
     3
                   4
                             1
                                    1
                   5
                            0
                                    3
```

5	6	0	3
6	7	0	1
7	8	0	3
8	9	1	3
9	10	1	2
10	11	1	3
11	12	1	1
12	13	0	3
13	14	0	3
14	15	0	3
15	16	1	2
16	17	0	3
17	18	1	2
18	19	0	3
19	20	1	3
20	21	0	2
21	22	1	2
22	23	1	3
23	24	1	1
24	25	0	3

	Name	Sex	Age	SibSp	\
0	Braund, Mr. Owen Harris	1	22.000000	1	
1	Cumings, Mrs. John Bradley (Florence Briggs Th	0 3	38.000000	1	
2	Heikkinen, Miss. Laina	0	26.000000	0	
3	Futrelle, Mrs. Jacques Heath (Lily May Peel)	0	35.000000	1	
4	Allen, Mr. William Henry	1	35.000000	0	
5	Moran, Mr. James	1	29.699118	0	
6	McCarthy, Mr. Timothy J	1	54.000000	0	
7	Palsson, Master. Gosta Leonard	1	2.000000	3	
8	Johnson, Mrs. Oscar W (Elisabeth Vilhelmina Berg)	0	27.000000	0	
9	Nasser, Mrs. Nicholas (Adele Achem)	0	14.000000	1	
10	Sandstrom, Miss. Marguerite Rut	0	4.000000	1	
11	Bonnell, Miss. Elizabeth	0	58.000000	0	
12	Saundercock, Mr. William Henry	1	20.000000	0	
13	Andersson, Mr. Anders Johan	1	39.000000	1	
14	Vestrom, Miss. Hulda Amanda Adolfina	0	14.000000	0	
15	Hewlett, Mrs. (Mary D Kingcome)	0	55.000000	0	
16	Rice, Master. Eugene	1	2.000000	4	
17	Williams, Mr. Charles Eugene	1	29.699118	0	
18	Vander Planke, Mrs. Julius (Emelia Maria Vande	0 3	31.000000	1	
19	Masselmani, Mrs. Fatima	0	29.699118	0	
20	Fynney, Mr. Joseph J	1	35.000000	0	
21	Beesley, Mr. Lawrence	1	34.000000	0	
22	McGowan, Miss. Anna "Annie"	0	15.000000	0	
23	Sloper, Mr. William Thompson	1	28.000000	0	
24	Palsson, Miss. Torborg Danira	0	8.000000	3	

	Parch	Ticket	Fare	Cabin	Embarked	Cabin_Letter
0	0	A/5 21171	7.2500	NaN	2	8
1	0	PC 17599	71.2833	C85	0	2
2	0	STON/02. 3101282	7.9250	NaN	2	8
3	0	113803	53.1000	C123	2	2
4	0	373450	8.0500	NaN	2	8
5	0	330877	8.4583	NaN	1	8
6	0	17463	51.8625	E46	2	4
7	1	349909	21.0750	NaN	2	8
8	2	347742	11.1333	NaN	2	8
9	0	237736	30.0708	NaN	0	8
10	1	PP 9549	16.7000	G6	2	6
11	0	113783	26.5500	C103	2	2
12	0	A/5. 2151	8.0500	NaN	2	8
13	5	347082	31.2750	NaN	2	8
14	0	350406	7.8542	NaN	2	8
15	0	248706	16.0000	NaN	2	8
16	1	382652	29.1250	NaN	1	8
17	0	244373	13.0000	NaN	2	8
18	0	345763	18.0000	NaN	2	8
19	0	2649	7.2250	NaN	0	8
20	0	239865	26.0000	NaN	2	8
21	0	248698	13.0000	D56	2	3
22	0	330923	8.0292	NaN	1	8
23	0	113788	35.5000	A6	2	0
24	1	349909	21.0750	NaN	2	8

[34]: train_set.corr(numeric_only=1)

[34]: PassengerId Survived Pclass Sex SibSp \ Age 1.000000 -0.005007 -0.035144 0.042939 0.033207 -0.057527 PassengerId Survived -0.005007 1.000000 -0.338481 -0.543351 -0.069809 -0.035322 Pclass -0.035144 -0.338481 1.000000 0.131900 -0.331339 0.042939 -0.543351 0.131900 Sex 1.000000 0.084153 -0.114631 Age $0.033207 \ -0.069809 \ -0.331339 \ \ 0.084153 \ \ 1.000000 \ -0.232625$ SibSp -0.057527 -0.035322 0.083081 -0.114631 -0.232625 1.000000 Parch -0.001652 0.081629 0.018443 -0.245489 -0.179191 0.414838 Fare 0.159651 Embarked 0.068230 Cabin_Letter -0.030939 -0.301116 0.746616 0.123076 -0.249134 0.041540

 Parch
 Fare
 Embarked
 Cabin_Letter

 PassengerId
 -0.001652
 0.012658
 0.013128
 -0.030939

 Survived
 0.081629
 0.257307
 -0.167675
 -0.301116

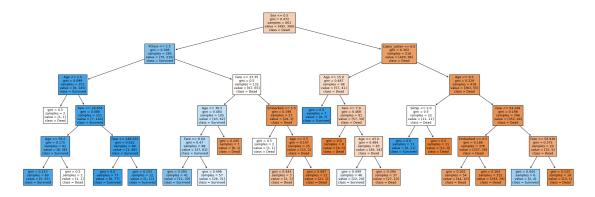
 Pclass
 0.018443
 -0.549500
 0.162098
 0.746616

 Sex
 -0.245489
 -0.182333
 0.108262
 0.123076

```
-0.179191 0.091566 -0.026749
     Age
                   0.414838 0.159651 0.068230
     SibSp
                                                    0.041540
     Parch
                   1.000000 0.216225 0.039798
                                                   -0.032548
                   0.216225 1.000000 -0.224719
     Fare
                                                   -0.523013
     Embarked
                   0.039798 -0.224719 1.000000
                                                    0.194255
     Cabin_Letter -0.032548 -0.523013 0.194255
                                                    1.000000
[35]: X = train_set[['Pclass', 'Sex', 'Age', 'SibSp', 'Parch', 'Fare', |
      # X = train_set[['Pclass', 'Sex', 'Age', 'SibSp', 'Parch', 'Fare', 'Embarked']]
     y = train set['Survived']
[36]: x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.1,_
      ⇔random_state=65)
[37]: from sklearn.metrics import accuracy_score
     from sklearn.tree import DecisionTreeClassifier
     from sklearn.model_selection import GridSearchCV
     param_grid = {
          'max_depth': [3, 5, 7, None],
          'min_samples_split': [2, 5, 10],
         'min_samples_leaf': [1, 2, 4]
     dt = DecisionTreeClassifier()
     gs = GridSearchCV(estimator=dt, param_grid=param_grid, scoring='accuracy', cv=8)
     gs.fit(x_train, y_train)
     best clf = gs.best estimator
     y_pred = best_clf.predict(x_test)
     yt_pred = best_clf.predict(x_train)
     test_accuracy = accuracy_score(y_test, y_pred)
     train_accuracy = accuracy_score(y_train, yt_pred)
     # Print the best parameters found
     print("Best Parameters:", gs.best_params_)
     print("Test Accuracy:", test_accuracy)
     print("Train Accuracy:", train_accuracy)
     #plot tree
     plot.figure(figsize=(30, 10))
     tree.plot_tree(best_clf, feature_names=list(X.columns), class_names=['Dead',__
      plot.show()
     Best Parameters: {'max_depth': 5, 'min_samples_leaf': 2, 'min_samples_split':
     10}
     Test Accuracy: 0.82222222222222
```

-0.249134

Train Accuracy: 0.846441947565543



```
[38]: from sklearn.model_selection import GridSearchCV
      from sklearn.ensemble import RandomForestClassifier
      rf = RandomForestClassifier(max_features='sqrt', oob_score=True,__
       ⇒random_state=65, n_jobs=-1)
      param_grid = { "criterion" : ["gini", "entropy"], "min_samples_leaf" : [1, 5, __
       410], "min_samples_split": [2, 4, 10, 12, 16], "n_estimators": [50, 100, ]
       →400, 700, 1000]}
      gs = GridSearchCV(estimator=rf, param_grid=param_grid, scoring='accuracy', u
       \hookrightarrowcv=3, n_jobs=-1)
      gs = gs.fit(x_train, y_train)
      print(gs.best score )
      print(gs.best_params_)
      print(gs.cv_results_)
     0.83270911360799
     {'criterion': 'entropy', 'min_samples_leaf': 1, 'min_samples_split': 10,
     'n estimators': 100}
     {'mean_fit_time': array([0.34907699, 0.47951063, 2.31736398, 3.98057111,
     6.09895651,
            0.36027638, 0.52474229, 2.27914437, 4.25119646, 4.66219838,
            0.21627617, 0.74940578, 2.28568403, 3.95806193, 5.10880907,
            0.34962797, 0.6259199 , 2.32556979, 3.81450311, 5.21995211,
            0.24948502, 0.472001, 2.41081119, 3.69699502, 5.40499751,
            0.27779945, 0.54392282, 2.37712789, 4.43397546, 5.43011244,
            0.29737163, 0.6869212 , 2.19706726, 4.30621847, 5.26877371,
            0.31758889, 0.69795338, 2.27575501, 4.13023901, 5.31245263,
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            0.2785267, 0.48835429, 2.52662849, 4.1319441, 5.78748178,
            0.41892974, 0.61710167, 2.29580863, 4.37137198, 5.01174005,
            0.27848403, 0.57237601, 1.91613801, 3.80944037, 5.09366051,
            0.27876981, 0.51791573, 2.13594246, 3.92401052, 5.43244489,
            0.29948306, 0.56669935, 2.12246052, 3.53022687, 4.82387249,
```

```
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       0.23433709, 0.54917614, 2.75081452, 5.85532451, 8.07800992,
       0.37149151, 1.08914002, 3.49035962, 5.00185378, 5.90075755,
       0.30375075, 1.19137732, 2.47537692, 4.99299844, 6.15784907,
       0.54188983, 0.80590494, 3.35590744, 5.73874728, 7.42789753,
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       0.32697988, 0.62200348, 2.40692266, 4.46010733, 5.71907465,
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       9.84429180e-02, 1.84025516e-01, 4.55329912e-01, 1.61673722e-01,
       1.76552052e-02, 3.47172241e-02, 4.70250318e-02, 3.50975611e-01,
       7.22511842e-02, 1.49320038e-02, 4.00566317e-03, 8.11429450e-02,
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       4.91200332e-01, 2.19703036e-01, 8.92604672e-02, 4.32002054e-02,
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       4.94933579e-02, 2.07145510e-01, 9.25125556e-02, 1.25637769e-02,
       5.73974246e-03, 1.65024439e-01, 2.90836797e-01, 4.15379203e-02,
       2.86660186e-02, 8.54558385e-02, 1.93403722e-01, 2.18034673e-01,
       3.99167500e-01, 5.82273063e-02, 2.01000079e-01, 2.40141249e-01,
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```

```
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[39]: gs = gs.fit(x_test, y_test)
print(gs.best_score_)
```

0.8222222222223

1 Predicting on test data

```
[53]: test_set.head(25)
```

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[53]:
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                                                                                 Name
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                   892
                   893
                              3
      1
                                                   Wilkes, Mrs. James (Ellen Needs)
      2
                              2
                                                          Myles, Mr. Thomas Francis
                   894
      3
                   895
                              3
                                                                    Wirz, Mr. Albert
      4
                   896
                              3
                                      Hirvonen, Mrs. Alexander (Helga E Lindqvist)
      5
                   897
                              3
                                                         Svensson, Mr. Johan Cervin
      6
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                              3
                                                                Connolly, Miss. Kate
      7
                              2
                                                       Caldwell, Mr. Albert Francis
                   899
      8
                   900
                              3
                                         Abrahim, Mrs. Joseph (Sophie Halaut Easu)
      9
                   901
                              3
                                                             Davies, Mr. John Samuel
                              3
      10
                   902
                                                                    Ilieff, Mr. Ylio
      11
                   903
                              1
                                                          Jones, Mr. Charles Cresson
                                     Snyder, Mrs. John Pillsbury (Nelle Stevenson)
      12
                   904
                              1
                              2
      13
                   905
                                                                Howard, Mr. Benjamin
      14
                   906
                                 Chaffee, Mrs. Herbert Fuller (Carrie Constance...
                              1
      15
                   907
                              2
                                     del Carlo, Mrs. Sebastiano (Argenia Genovesi)
```

16		908	2			Kea	ne, Mr. Daniel		
17		909	3			Ass	af, Mr. Gerios		
18		910	3	Ilmakangas, Miss. Ida Livija					
19		911	3	Assaf Khalil, Mrs. Mariana (Miriam")"					
20		912	1			Rothschi	ld, Mr. Martin		
21		913	3		0	lsen, Mast	er. Artur Karl		
22		914	1		Flegenheim,	Mrs. Alfre	d (Antoinette)		
23		915	1		~		hard Norris II		
24		916	1	Ryers	on, Mrs. Arthur La				
				v					
	Sex	Age	SibSp	Parch	Ticket	Fare	Cabir	ı \	
0	1	34.50000	0	0	330911	7.8292	NaN	Ī	
1	0	47.00000	1	0	363272	7.0000	NaN	Ī	
2	1	62.00000	0	0	240276	9.6875	NaN	Ī	
3	1	27.00000	0	0	315154	8.6625	NaN	Ī	
4	0	22.00000	1	1	3101298	12.2875	NaN	Ī	
5	1	14.00000	0	0	7538	9.2250	NaN	Ī	
6	0	30.00000	0	0	330972	7.6292	NaN	Ī	
7	1	26.00000	1	1	248738	29.0000	NaN	Ī	
8	0	18.00000	0	0	2657	7.2292	NaN	Ī	
9	1	21.00000	2	0	A/4 48871	24.1500	NaN	Ī	
10	1	30.27259	0	0	349220	7.8958	NaN	Ī	
11	1	46.00000	0	0	694	26.0000	NaN	Ī	
12	0	23.00000	1	0	21228	82.2667	B45		
13	1	63.00000	1	0	24065	26.0000	NaN	Ī	
14	0	47.00000	1	0	W.E.P. 5734	61.1750	E31		
15	0	24.00000	1	0	SC/PARIS 2167	27.7208	NaN	Ī	
16	1	35.00000	0	0	233734	12.3500	NaN	Ī	
17	1	21.00000	0	0	2692	7.2250	NaN	Ī	
18	0	27.00000	1	0	STON/02. 3101270	7.9250	NaN	Ī	
19	0	45.00000	0	0	2696	7.2250	NaN	Ī	
20	1	55.00000	1	0	PC 17603	59.4000	NaN	Ī	
21	1	9.00000	0	1	C 17368	3.1708	NaN	Ī	
22	0	30.27259	0	0	PC 17598	31.6833	NaN	Ī	
23	1	21.00000	0	1	PC 17597	61.3792	NaN	Ī	
24	0	48.00000	1	3	PC 17608	262.3750	B57 B59 B63 B66	;	
	Emba	rked Cabi	n_Lette	r					
0		1	•	7					
1		2	•	7					
2		1	•	7					
3		2	•	7					
4		2	•	7					
5		2	•	7					
6		1	-	7					
7		2	•	7					
8		0	•	7					

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10
                 2
                               7
                2
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      11
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      12
                               1
      13
                2
                               7
                2
      14
                               4
      15
                0
                               7
      16
                 1
                               7
                               7
      17
                0
      18
                2
                               7
                               7
      19
                0
     20
                0
                               7
                2
                               7
      21
      22
                2
                               7
      23
                 0
                               7
      24
                 0
                               1
[54]: test_set.isnull().sum()
[54]: PassengerId
                       0
     Pclass
                       0
     Name
                       0
     Sex
                       0
                       0
      Age
      SibSp
                       0
                       0
     Parch
                       0
     Ticket
     Fare
                       0
      Cabin
                      327
      Embarked
                       0
      Cabin_Letter
                       0
      dtype: int64
[58]: test_X = test_set[['Pclass', 'Sex', 'Age', 'SibSp', 'Parch', __
      # Make predictions on the test data using the trained Decision Tree classifier
      test_predictions_dt = best_clf.predict(test_X)
      # Make predictions on the test data using the trained Random Forest classifier
      test_predictions_rf = gs.predict(test_X)
      # Output predictions
      output_dt = pnd.DataFrame({'PassengerId': test_set.PassengerId, 'Survived':
      →test_predictions_dt})
      output_rf = pnd.DataFrame({'PassengerId': test_set.PassengerId, 'Survived':
      →test_predictions_rf})
      # Save the predictions to a CSV file
      output_dt.to_csv('predictions_decision_tree.csv', index=False)
      output_rf.to_csv('predictions_random_forest.csv', index=False)
```

[60]: compareDtAndRf = accuracy_score(test_predictions_dt, test_predictions_rf)
print("Comparing results predicted by Decision Tree and Random Forest",

→compareDtAndRf)

Comparing results predicted by Decision Tree and Random Forest 0.8444976076555024

[]: