

Titanic Classification

September 13, 2023

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
[2]: titanic_data = pd.read_csv('C:/Users/ASUS/Downloads/Titanic/train.csv')
```

```
[3]: titanic_data
```

```
[3]:      PassengerId  Survived  Pclass \
0                1         0        3
1                2         1        1
2                3         1        3
3                4         1        1
4                5         0        3
..          ...
886            887         0        2
887            888         1        1
888            889         0        3
889            890         1        1
890            891         0        3
```

```
      Name      Sex  Age  SibSp \
0  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    male  35.0    0
..          ...
886  Montvila, Rev. Juozas    male  27.0    0
887  Graham, Miss. Margaret Edith    female  19.0    0
888  Johnston, Miss. Catherine Helen "Carrie"    female   NaN    1
889  Behr, Mr. Karl Howell    male  26.0    0
890  Dooley, Mr. Patrick    male  32.0    0
```

```
      Parch      Ticket    Fare Cabin Embarked
0         0      A/5 21171   7.2500   NaN        S
1         0      PC 17599  71.2833   C85        C
2         0  STON/O2. 3101282   7.9250   NaN        S
```

3	0	113803	53.1000	C123	S
4	0	373450	8.0500	NaN	S
..
886	0	211536	13.0000	NaN	S
887	0	112053	30.0000	B42	S
888	2	W./C. 6607	23.4500	NaN	S
889	0	111369	30.0000	C148	C
890	0	370376	7.7500	NaN	Q

[891 rows x 12 columns]

```
[4]: titanic_data.head()
```

```
[4]: PassengerId  Survived  Pclass  \
0            1         0         3
1            2         1         1
2            3         1         3
3            4         1         1
4            5         0         3

                                     Name    Sex  Age  SibSp  \
0                               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    male  35.0      0

   Parch    Ticket   Fare Cabin Embarked
0      0   A/5 21171   7.2500   NaN        S
1      0   PC 17599  71.2833   C85        C
2      0 STON/O2. 3101282   7.9250   NaN        S
3      0   113803  53.1000  C123        S
4      0   373450   8.0500   NaN        S
```

```
[5]: titanic_data.describe()
```

```
[5]: PassengerId  Survived  Pclass    Age  SibSp  \
count  891.000000  891.000000  891.000000  714.000000  891.000000
mean    446.000000    0.383838    2.308642   29.699118    0.523008
std    257.353842    0.486592    0.836071   14.526497    1.102743
min      1.000000    0.000000    1.000000    0.420000    0.000000
25%    223.500000    0.000000    2.000000   20.125000    0.000000
50%    446.000000    0.000000    3.000000   28.000000    0.000000
75%    668.500000    1.000000    3.000000   38.000000    1.000000
max    891.000000    1.000000    3.000000   80.000000    8.000000

      Parch    Fare
count  891.000000  891.000000
mean    1.000000   54.181136
std     1.000000   53.901141
min     0.000000    5.000000
25%     0.000000   21.000000
50%     1.000000   53.000000
75%     2.000000   51.000000
max     4.000000  512.000000
```

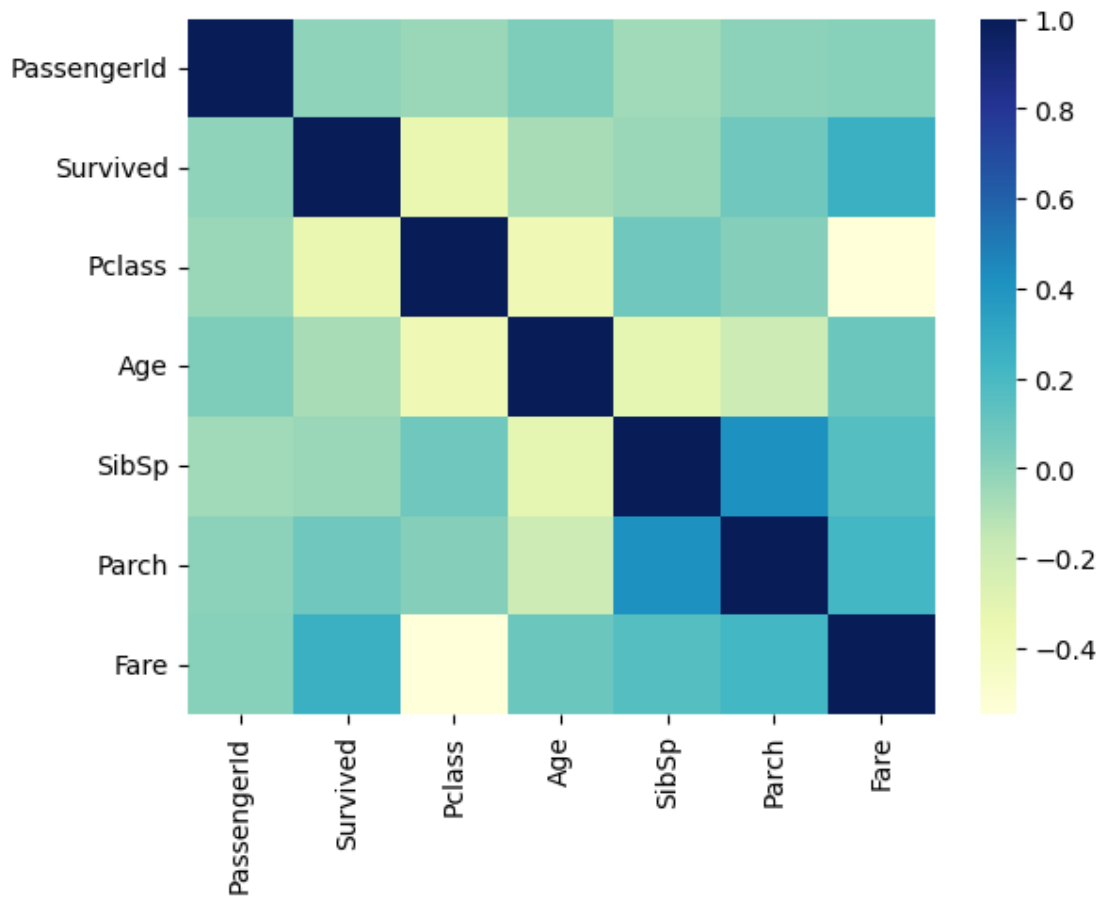
count	891.000000	891.000000
mean	0.381594	32.204208
std	0.806057	49.693429
min	0.000000	0.000000
25%	0.000000	7.910400
50%	0.000000	14.454200
75%	0.000000	31.000000
max	6.000000	512.329200

```
[6]: import seaborn as sns

sns.heatmap(titanic_data.corr(), cmap="YlGnBu")
plt.show()
```

C:\Users\ASUS\AppData\Local\Temp\ipykernel_26076\1602845089.py:3: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.

```
sns.heatmap(titanic_data.corr(), cmap="YlGnBu")
```



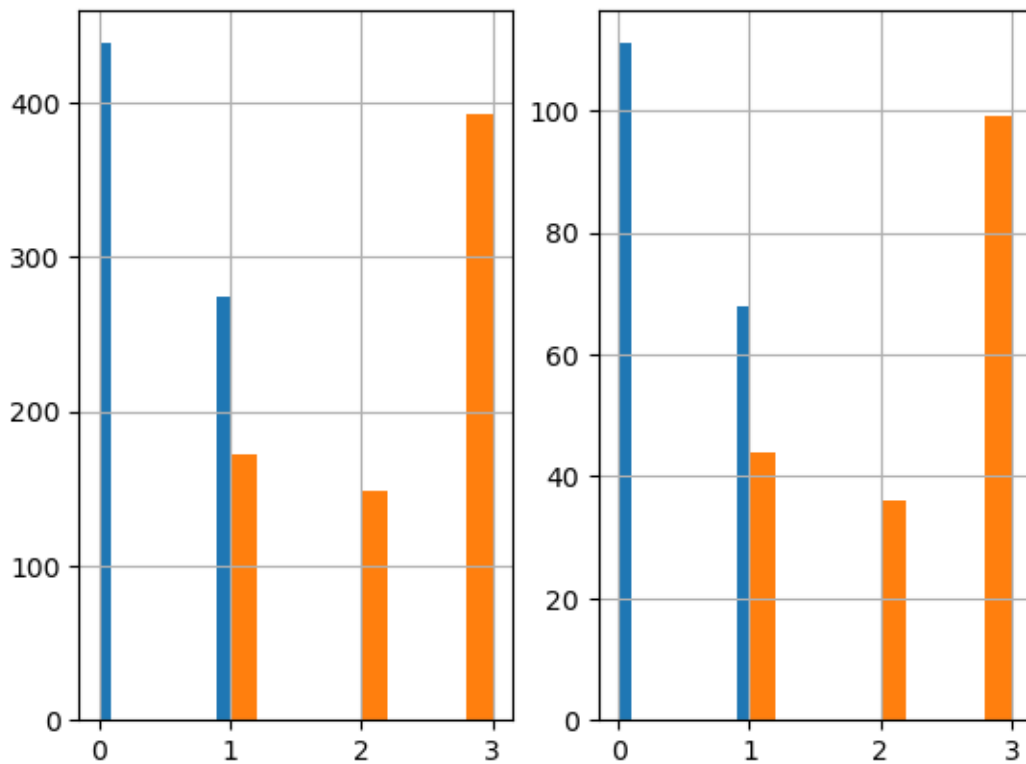
```
[7]: from sklearn.model_selection import StratifiedShuffleSplit

split = StratifiedShuffleSplit(n_splits=1, test_size=0.2)
for train_indices, test_indices in split.split(titanic_data,
→titanic_data[["Survived", "Pclass", "Sex"]]):
    strat_train_set = titanic_data.loc[train_indices]
    strat_test_set = titanic_data.loc[test_indices]
```

```
[8]: plt.subplot(1,2,1)
strat_train_set['Survived'].hist()
strat_train_set['Pclass'].hist()

plt.subplot(1,2,2)
strat_test_set['Survived'].hist()
strat_test_set['Pclass'].hist()

plt.show()
```



```
[9]: strat_train_set.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 712 entries, 362 to 16
```

Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	PassengerId	712 non-null	int64
1	Survived	712 non-null	int64
2	Pclass	712 non-null	int64
3	Name	712 non-null	object
4	Sex	712 non-null	object
5	Age	573 non-null	float64
6	SibSp	712 non-null	int64
7	Parch	712 non-null	int64
8	Ticket	712 non-null	object
9	Fare	712 non-null	float64
10	Cabin	157 non-null	object
11	Embarked	711 non-null	object

dtypes: float64(2), int64(5), object(5)

memory usage: 72.3+ KB

```
[10]: from sklearn.base import BaseEstimator, TransformerMixin
      from sklearn.impute import SimpleImputer
```

```
class AgeImputer(BaseEstimator, TransformerMixin):
```

```
    def fit(self, x, y=None):
        return self
```

```
    def transform(self, X):
        imputer = SimpleImputer(strategy="mean")
        X['Age']=imputer.fit_transform(X[["Age"]])
        return X
```

```
[11]: from sklearn.preprocessing import OneHotEncoder
```

```
class FeatureEncoder(BaseEstimator, TransformerMixin):
```

```
    def fit(self, X):
        return self
```

```
    def transform(self, X):
        encoder = OneHotEncoder()
        matrix = encoder.fit_transform(X[["Embarked"]]).toarray()
```

```
        column_names = ["C", "S", "Q", "N"]
```

```
        for i in range(len(matrix.T)):
            X[column_names[i]] = matrix.T[i]
```

```

matrix = encoder.fit_transform(X[['Sex']]).toarray()

column_names = ["Female", "Male"]

for i in range(len(matrix.T)):
    X[column_names[i]] = matrix.T[i]

return X

```

```

[12]: class FeatureDropper(BaseEstimator, TransformerMixin):

    def fit(self, X, y = None):
        return self

    def transform(self, X):
        return X.drop(["Embarked", "Name", "Ticket", "Cabin", "Sex", "N"], axis=
↪ 1, errors="ignore")

```

```

[13]: from sklearn.pipeline import Pipeline

pipeline = Pipeline([("ageimputer", AgeImputer()),
                      ("featureencoder", FeatureEncoder()),
                      ("featuredropper", FeatureDropper())])

```

```

[14]: strat_train_set = pipeline.fit_transform(strat_train_set)

```

```

[15]: strat_train_set

```

```

[15]: PassengerId  Survived  Pclass    Age  SibSp  Parch    Fare   C  \
362           363         0      3  45.000000    0      1   14.4542  1.0
532           533         0      3  17.000000    1      1    7.2292  1.0
699           700         0      3  42.000000    0      0    7.6500  0.0
160           161         0      3  44.000000    0      1   16.1000  0.0
828           829         1      3  29.585951    0      0    7.7500  0.0
..           ...         ...     ...     ...     ...     ...     ...
507           508         1      1  29.585951    0      0   26.5500  0.0
805           806         0      3  31.000000    0      0    7.7750  0.0
181           182         0      2  29.585951    0      0   15.0500  1.0
737           738         1      1  35.000000    0      0  512.3292  1.0
16             17         0      3   2.000000    4      1   29.1250  0.0

      S    Q  Female  Male
362  0.0  0.0     1.0   0.0
532  0.0  0.0     0.0   1.0
699  0.0  1.0     0.0   1.0
160  0.0  1.0     0.0   1.0
828  1.0  0.0     0.0   1.0

```

```

..    ...    ...    ...    ...
507  0.0  1.0    0.0  1.0
805  0.0  1.0    0.0  1.0
181  0.0  0.0    0.0  1.0
737  0.0  0.0    0.0  1.0
16   1.0  0.0    0.0  1.0

```

[712 rows x 12 columns]

```
[17]: strat_train_set.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 712 entries, 362 to 16
Data columns (total 12 columns):
#   Column          Non-Null Count  Dtype
---  -
0   PassengerId      712 non-null    int64
1   Survived         712 non-null    int64
2   Pclass           712 non-null    int64
3   Age              712 non-null    float64
4   SibSp            712 non-null    int64
5   Parch            712 non-null    int64
6   Fare             712 non-null    float64
7   C                 712 non-null    float64
8   S                 712 non-null    float64
9   Q                 712 non-null    float64
10  Female           712 non-null    float64
11  Male             712 non-null    float64
dtypes: float64(7), int64(5)
memory usage: 72.3 KB

```

```
[21]: from sklearn.preprocessing import StandardScaler
```

```

X = strat_train_set.drop(['Survived'], axis=1)
Y = strat_train_set['Survived']

scaler = StandardScaler()
X_data = scaler.fit_transform(X)
Y_data = Y.to_numpy()

```

```
[22]: X_data
```

```

[22]: array([[ -3.45543672e-01,  8.27893418e-01,  1.18806281e+00, ...,
        -1.63985340e+00,  1.35941164e+00, -1.35941164e+00],
       [ 3.23200629e-01,  8.27893418e-01, -9.70082596e-01, ...,
        -1.63985340e+00, -7.35612358e-01,  7.35612358e-01],
       [ 9.80143559e-01,  8.27893418e-01,  9.56832949e-01, ...,
        6.09810609e-01, -7.35612358e-01,  7.35612358e-01],

```

```
...,
[-1.05755966e+00, -3.70196244e-01, 2.73831169e-16, ...,
-1.63985340e+00, -7.35612358e-01, 7.35612358e-01],
[ 1.12962758e+00, -1.56828591e+00, 4.17296597e-01, ...,
-1.63985340e+00, -7.35612358e-01, 7.35612358e-01],
[-1.70663501e+00, 8.27893418e-01, -2.12623192e+00, ...,
-1.63985340e+00, -7.35612358e-01, 7.35612358e-01]])
```

```
[23]: Y_data
```

```
[23]: array([0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0,
0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 0,
0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 0, 1, 0, 1,
1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0,
1, 1, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1,
0, 0, 1, 0, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1,
1, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1,
1, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1,
1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1,
0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 0,
1, 0, 1, 1, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1, 1, 0, 0, 0,
0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1,
1, 1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 0,
1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 1,
1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, 0,
0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 0, 1,
1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0,
0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0,
0, 0, 1, 0, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0,
0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 1,
0, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0,
0, 1, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 1,
1, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1,
0, 0, 0, 1, 0, 0, 1, 0], dtype=int64)
```

```
[27]: from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import GridSearchCV
```



```

clf = RandomForestClassifier()

param_gird = [
    {"n_estimators": [10,100,200,500], "max_depth": [None,5,10],
    ↪ "min_samples_split": [2,3,4]}
]

grid_search = GridSearchCV(clf, param_gird, cv=3, scoring="accuracy",
    ↪ return_train_score=True)
grid_search.fit(X_data, Y_data)

```

```

[27]: GridSearchCV(cv=3, estimator=RandomForestClassifier(),
    param_grid=[{'max_depth': [None, 5, 10],
    'min_samples_split': [2, 3, 4],
    'n_estimators': [10, 100, 200, 500]}],
    return_train_score=True, scoring='accuracy')

```

```

[28]: final_clf = grid_search.best_estimator_

```

```

[29]: final_clf

```

```

[29]: RandomForestClassifier(max_depth=5, min_samples_split=4)

```

```

[30]: strat_test_set = pipeline.fit_transform(strat_test_set)

```

```

[31]: strat_test_set

```

```

[31]:
    PassengerId  Survived  Pclass    Age  SibSp  Parch    Fare   C  \
769          770         0      3  32.000000     0     0    8.3625  0.0
857          858         1      1  51.000000     0     0   26.5500  0.0
63           64         0      3   4.000000     3     2   27.9000  0.0
872          873         0      1  33.000000     0     0    5.0000  0.0
15           16         1      2  55.000000     0     0   16.0000  0.0
..          ...         ...     ...     ...     ...     ...     ...
880          881         1      2  25.000000     0     1   26.0000  0.0
716          717         1      1  38.000000     0     0  227.5250  1.0
711          712         0      1  30.159007     0     0   26.5500  0.0
326          327         0      3  61.000000     0     0    6.2375  0.0
462          463         0      1  47.000000     0     0   38.5000  0.0

      S    Q  Female  Male
769  0.0  1.0     0.0   1.0
857  0.0  1.0     0.0   1.0
63   0.0  1.0     0.0   1.0
872  0.0  1.0     0.0   1.0
15   0.0  1.0     1.0   0.0
..   ...  ...     ...   ...

```

```

880  0.0  1.0    1.0  0.0
716  0.0  0.0    1.0  0.0
711  0.0  1.0    0.0  1.0
326  0.0  1.0    0.0  1.0
462  0.0  1.0    0.0  1.0

```

[179 rows x 12 columns]

```

[35]: X_test = strat_test_set.drop(['Survived'], axis=1)
      Y_test = strat_test_set['Survived']

      scaler = StandardScaler()
      X_data_test = scaler.fit_transform(X_test)
      Y_data_test = Y_test.to_numpy()

```

```

[36]: final_clf.score(X_data_test, Y_data_test)

```

```

[36]: 0.8156424581005587

```

```

[37]: final_data = pipeline.fit_transform(titanic_data)

```

```

[38]: final_data

```

```

[38]: PassengerId  Survived  Pclass     Age  SibSp  Parch    Fare   C  \
0             1         0       3  22.000000     1     0    7.2500  0.0
1             2         1       1  38.000000     1     0   71.2833  1.0
2             3         1       3  26.000000     0     0    7.9250  0.0
3             4         1       1  35.000000     1     0   53.1000  0.0
4             5         0       3  35.000000     0     0    8.0500  0.0
..         ...         ...     ...     ...     ...     ...     ...
886          887         0       2  27.000000     0     0   13.0000  0.0
887          888         1       1  19.000000     0     0   30.0000  0.0
888          889         0       3  29.699118     1     2   23.4500  0.0
889          890         1       1  26.000000     0     0   30.0000  1.0
890          891         0       3  32.000000     0     0    7.7500  0.0

```

```

      S    Q  Female  Male
0    0.0  1.0     0.0   1.0
1    0.0  0.0     1.0   0.0
2    0.0  1.0     1.0   0.0
3    0.0  1.0     1.0   0.0
4    0.0  1.0     0.0   1.0
..    ...  ...     ...   ...
886  0.0  1.0     0.0   1.0
887  0.0  1.0     1.0   0.0
888  0.0  1.0     1.0   0.0
889  0.0  0.0     0.0   1.0

```

```
890  1.0  0.0    0.0  1.0
```

```
[891 rows x 12 columns]
```

```
[41]: X_final = final_data.drop(['Survived'], axis=1)
      Y_final = final_data['Survived']
```

```
scaler = StandardScaler()
X_data_final = scaler.fit_transform(X_final)
Y_data_final = Y_final.to_numpy()
```

```
[42]: prod_clf = RandomForestClassifier()

param_gird = [
    {'n_estimators': [10,100,200,500], "max_depth": [None,5,10],
    ↪ "min_samples_split": [2,3,4]}
]

grid_search = GridSearchCV(prod_clf, param_gird, cv=3, scoring="accuracy",
    ↪ return_train_score=True)
grid_search.fit(X_data_final, Y_data_final)
```

```
[42]: GridSearchCV(cv=3, estimator=RandomForestClassifier(),
                  param_grid=[{'max_depth': [None, 5, 10],
                              'min_samples_split': [2, 3, 4],
                              'n_estimators': [10, 100, 200, 500]}],
                  return_train_score=True, scoring='accuracy')
```

```
[43]: prod_final_clf = grid_search.best_estimator_
```

```
[45]: prod_final_clf
```

```
[45]: RandomForestClassifier(max_depth=5, min_samples_split=3, n_estimators=200)
```

```
[46]: titanic_test_data = pd.read_csv('C:/Users/ASUS/Downloads/Titanic/test.csv')
```

```
[48]: titanic_test_data
```

```
[48]:
```

	PassengerId	Pclass	Name \
0	892	3	Kelly, Mr. James
1	893	3	Wilkes, Mrs. James (Ellen Needs)
2	894	2	Myles, Mr. Thomas Francis
3	895	3	Wirz, Mr. Albert
4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)
..
413	1305	3	Spector, Mr. Woolf

414	1306	1	Oliva y Ocana, Dona. Fermina
415	1307	3	Saether, Mr. Simon Sivertsen
416	1308	3	Ware, Mr. Frederick
417	1309	3	Peter, Master. Michael J

	Sex	Age	SibSp	Parch		Ticket	Fare	Cabin	Embarked
0	male	34.5	0	0		330911	7.8292	NaN	Q
1	female	47.0	1	0		363272	7.0000	NaN	S
2	male	62.0	0	0		240276	9.6875	NaN	Q
3	male	27.0	0	0		315154	8.6625	NaN	S
4	female	22.0	1	1		3101298	12.2875	NaN	S
..
413	male	NaN	0	0		A.5. 3236	8.0500	NaN	S
414	female	39.0	0	0		PC 17758	108.9000	C105	C
415	male	38.5	0	0	SOTON/O.Q.	3101262	7.2500	NaN	S
416	male	NaN	0	0		359309	8.0500	NaN	S
417	male	NaN	1	1		2668	22.3583	NaN	C

[418 rows x 11 columns]

```
[49]: final_test_data = pipeline.fit_transform(titanic_test_data)
```

```
[50]: final_test_data
```

```
[50]:
```

	PassengerId	Pclass	Age	SibSp	Parch	Fare	C	S	Q	\
0	892	3	34.50000	0	0	7.8292	0.0	1.0	0.0	
1	893	3	47.00000	1	0	7.0000	0.0	0.0	1.0	
2	894	2	62.00000	0	0	9.6875	0.0	1.0	0.0	
3	895	3	27.00000	0	0	8.6625	0.0	0.0	1.0	
4	896	3	22.00000	1	1	12.2875	0.0	0.0	1.0	
..	
413	1305	3	30.27259	0	0	8.0500	0.0	0.0	1.0	
414	1306	1	39.00000	0	0	108.9000	1.0	0.0	0.0	
415	1307	3	38.50000	0	0	7.2500	0.0	0.0	1.0	
416	1308	3	30.27259	0	0	8.0500	0.0	0.0	1.0	
417	1309	3	30.27259	1	1	22.3583	1.0	0.0	0.0	

	Female	Male
0	0.0	1.0
1	1.0	0.0
2	0.0	1.0
3	0.0	1.0
4	1.0	0.0
..
413	0.0	1.0
414	1.0	0.0
415	0.0	1.0

```
416      0.0    1.0
417      0.0    1.0
```

```
[418 rows x 11 columns]
```

```
[51]: final_test_data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 418 entries, 0 to 417
Data columns (total 11 columns):
#   Column      Non-Null Count  Dtype
---  -
0   PassengerId  418 non-null    int64
1   Pclass       418 non-null    int64
2   Age          418 non-null    float64
3   SibSp        418 non-null    int64
4   Parch        418 non-null    int64
5   Fare         417 non-null    float64
6   C            418 non-null    float64
7   S            418 non-null    float64
8   Q            418 non-null    float64
9   Female       418 non-null    float64
10  Male         418 non-null    float64
dtypes: float64(7), int64(4)
memory usage: 36.1 KB
```

```
[55]: X_final_test = final_test_data
X_final_test = X_final_test.fillna(method="ffill")

scaler = StandardScaler()
X_data_final_test = scaler.fit_transform(X_final_test)
```

```
[56]: predictions = prod_final_clf.predict(X_data_final_test)
```

```
[57]: predictions
```

```
[57]: array([0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0,
        1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1,
        1, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1,
        1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1,
        1, 1, 1, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0,
        0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
        0, 0, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1,
        1, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1,
        0, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0,
        1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1, 1,
        0, 0, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1,
```

```

0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0,
0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1,
0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0,
1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0,
0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0,
1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1,
0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0],
dtype=int64)

```

```

[58]: final_df = pd.DataFrame(titanic_test_data['PassengerId'])
final_df['Survived'] = predictions
final_df.to_csv('C:/Users/ASUS/Downloads/Titanic/predictions.csv', index=False)

```

```

[59]: final_df

```

```

[59]:
   PassengerId  Survived
0          892         0
1          893         0
2          894         0
3          895         0
4          896         1
...         ...      ...
413        1305         0
414        1306         1
415        1307         0
416        1308         0
417        1309         0

```

```

[418 rows x 2 columns]

```