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

In [2]: titanic_df = pd.read_csv("train.csv")

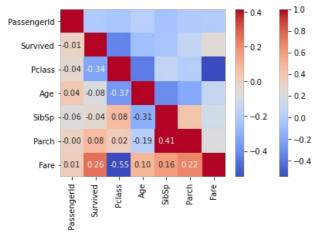
In [3]: titanic_df.head(5)
```

3]:		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
	0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
	1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	С
	2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
	3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
	4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S

```
In [4]: titanic_df.describe()
```

Out[3

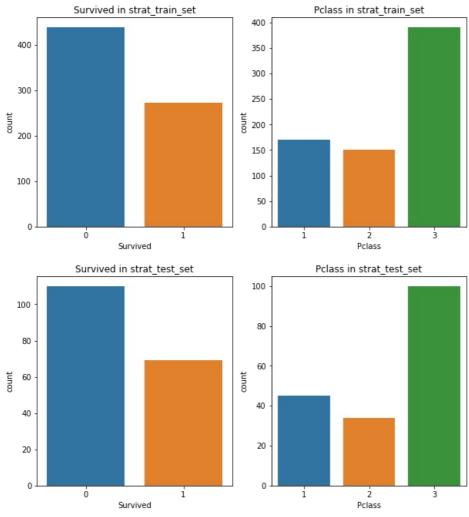
Out[4]:		Passengerld	Survived	Pclass	Age	SibSp	Parch	Fare
	count	891.000000	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000
	mean	446.000000	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208
	std	257.353842	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429
	min	1.000000	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000
	25%	223.500000	0.000000	2.000000	20.125000	0.000000	0.000000	7.910400
	50%	446.000000	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200
	75%	668.500000	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000
	max	891.000000	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200



```
# Create a StratifiedShuffleSplit object
split = StratifiedShuffleSplit(n_splits=1, test_size=0.2, random_state=42)

# Loop through the splits (we only have one)
for train_index, test_index in split.split(titanic_df, titanic_df['Survived'],titanic_df['Pclass']):
    # Get the training and testing data
    strat_train_set = titanic_df.iloc[train_index]
    strat_test_set = titanic_df.iloc[test_index]
```

```
In [7]:
         # Create a figure and axes object using plt.subplots
         fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(10, 5))
         # Use seaborn to create a bar plot of the 'Survived' column on the first axes
         sns.countplot(x='Survived', data=strat_train_set, ax=ax1)
ax1.set_title('Survived in strat_train_set')
         # Use seaborn to create a bar plot of the 'Pclass' column on the second axes
         sns.countplot(x='Pclass', data=strat train set, ax=ax2)
         ax2.set title('Pclass in strat train set')
         # Create a figure and axes object using plt.subplots
         fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(10, 5))
         # Use seaborn to create a bar plot of the 'Survived' column on the first axes
         sns.countplot(x='Survived', data=strat_test_set, ax=ax1)
         ax1.set_title('Survived in strat_test_set')
         # Use seaborn to create a bar plot of the 'Pclass' column on the second axes
         sns.countplot(x='Pclass', data=strat_test_set, ax=ax2)
         ax2.set title('Pclass in strat test set')
         plt.show()
```



```
In [8]: strat_train_set.info()
```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 712 entries, 692 to 507
Data columns (total 12 columns):
Column Non-Null Count Dtype

```
712 non-null
                                            obiect
              Sex
           5
              Age
                           575 non-null
                                            float64
              SibSp
                            712 non-null
                                             int64
                            712 non-null
              Parch
                                             int64
                           712 non-null
             Ticket
                                             object
                            712 non-null
          9
              Fare
                                             float64
                            160 non-null
           10 Cabin
                                             object
          11 Embarked
                            710 non-null
                                             object
         dtypes: float64(2), int64(5), object(5)
         memory usage: 88.5+ KB
 In [9]:
          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
In [10]:
          from sklearn.preprocessing import OneHotEncoder
          class FeatureEncoder (BaseEstimator, TransformerMixin):
               def fit (self, X, y=None) :
                   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
In [11]:
          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")
In [12]:
          class FeatureDropper(BaseEstimator, TransformerMixin):
              def __init__(self, columns_to_drop):
                   self.columns_to_drop = columns_to_drop
              def fit(self, X, y=None):
                   return self
              def transform(self, X):
                   return X.drop(self.columns_to_drop, axis=1, errors="ignore")
In [13]:
          fd = FeatureDropper(columns to drop = ["Embarked", "Name", "Ticket", "Cabin", "Sex", "N"])
In [14]:
          from sklearn.pipeline import Pipeline
           pipeline = Pipeline([("ageimputer", AgeImputer()),
                                  ("featureencoder", FeatureEncoder()),
("featuredropper", fd)])
In [15]:
          strat_train_set = pipeline.fit_transform(strat_train_set)
          <ipython-input-9-bf47ab59dc18>:11: SettingWithCopyWarning:
          A value is trying to be set on a copy of a slice from a DataFrame.
```

PassengerId 712 non-null

Survived 712 non-null

712 non-null

712 non-null

1

3

Pclass

Name

int64 int64

int64

object

```
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#retur
ning-a-view-versus-a-copy
 X['Age'] = imputer.fit transform(X[['Age']])
<ipython-input-10-9b26102771b5>:12: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#retur
ning-a-view-versus-a-copy
  X[column_names[i]] = matrix.T[i]
<ipython-input-10-9b26102771b5>:16: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#retur
ning-a-view-versus-a-copy
X[column_names[i]] = matrix.T[i]
```

In [16]:

strat train set

Age SibSp Parch Out[16]: Passengerld Survived Pclass Fare С s Q Female Male 692 693 3 29.807687 56.4958 0.0 0.0 1.0 0.0 1.0 481 482 29.807687 0 0.0000 0.0 0.0 1.0 0.0 1.0 527 528 0 29.807687 0 0.0 0.0 1.0 1.0 221.7792 0.0 855 856 18.000000 0 9.3500 0.0 0.0 1.0 1.0 0.0 801 802 31.000000 26.2500 0.0 0.0 1.0 1.0 0.0 359 360 1 3 29.807687 0 0 7.8792 0.0 1.0 0.0 1.0 0.0 258 259 35.000000 512.3292 1.0 0.0 0.0 1.0 0.0 736 737 0 48.000000 34.3750 0.0 0.0 1.0 1.0 0.0 462 463 0 47.000000 0 38.5000 0.0 0.0 1.0 0.0 1.0 507 508 1 29.807687 26.5500 0.0 0.0 1.0 0.0 1.0

712 rows × 12 columns

In [17]:

strat_train_set.info()

<class 'pandas.core.frame.DataFrame'> Int64Index: 712 entries, 692 to 507 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: 88.5 KB

```
In [18]:
          from sklearn.preprocessing import StandardScaler
          X = strat_train_set.drop(['Survived'], axis =1)
          y = strat_train_set['Survived']
          scaler = StandardScaler().fit(X)
          X data = scaler.transform(X)
          y data = y.to numpy()
```

```
In [19]: | 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)
Out[19]: 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')
In [20]:
          final clf = grid search.best estimator
In [21]:
          final clf
Out[21]: RandomForestClassifier(max depth=10, min samples split=4, n estimators=10)
In [22]:
          strat test set = pipeline.fit transform(strat test set)
         <ipython-input-9-bf47ab59dc18>:11: SettingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#retur
         ning-a-view-versus-a-copy
           X['Age'] = imputer.fit_transform(X[['Age']])
         <ipython-input-10-9b26102771b5>:12: SettingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#retur
         ning-a-view-versus-a-copy
           X[column_names[i]] = matrix.T[i]
         <ipython-input-10-9b26102771b5>:16: SettingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row indexer,col indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#retur
         ning-a-view-versus-a-copy
         X[column names[i]] = matrix.T[i]
In [23]:
          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()
In [24]:
          final clf.score(X data test, y data test)
Out[24]: 0.7877094972067039
In [25]:
          final data = pipeline.fit transform(titanic df)
In [26]:
          final data
                                            Age SibSp Parch
             Passengerld Survived Pclass
                                                              Fare
                                                                    С
                                                                        S
                                                                           Q Female Male
Out[26]:
           0
                      1
                              0
                                     3 22.000000
                                                            7.2500 0.0 0.0 1.0
                                                                                       1.0
                                                         0
                                                                                  0.0
                                     1 38.000000
                                                         0 71.2833 1.0 0.0 0.0
                                                                                  1.0
                                                                                       0.0
```

```
2
               3
                                  3 26.000000
                                                                7.9250 0.0 0.0 1.0
                                                                                          1.0
                                                                                                0.0
                                                                                                0.0
 3
                                     35.000000
                                                            0 53.1000 0.0 0.0 1.0
                                                                                          1.0
  4
               5
                         0
                                     35.000000
                                                    0
                                                                8.0500
                                                                       0.0
                                                                           0.0
                                                                                 1.0
                                                                                          0.0
                                                                                                1.0
 ...
886
             887
                         0
                                                    0
                                                            0 13.0000 0.0 0.0 1.0
                                  2 27.000000
                                                                                          0.0
                                                                                                1.0
887
             888
                                     19.000000
                                                    0
                                                              30.0000
                                                                       0.0
                                                                           0.0
                                                                                 1.0
                                                                                          1.0
                                                                                                0.0
888
             889
                                     29.699118
                                                               23.4500
                                                                       0.0
                                                                           0.0
                                                                                 1.0
                                                                                          1.0
                                                                                                0.0
889
             890
                                     26.000000
                                                    0
                                                              30.0000 1.0 0.0 0.0
                                                                                          0.0
                                                                                                1.0
890
             891
                         0
                                  3 32.000000
                                                    0
                                                                7.7500 0.0 1.0 0.0
                                                                                          0.0
                                                                                                1.0
```

891 rows × 12 columns

```
In [29]:
           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()
In [30]:
           production clf = RandomForestClassifier()
           param_gird =[
               {"n estimators": [10,100,200,500], "max depth": [None, 5, 10], "min samples split": [2,3,4]}
           grid search = GridSearchCV(production clf, param gird, cv=3, scoring = "accuracy", return train score=True)
           grid_search.fit(X_data_final, y_data_final)
Out[30]: 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')
In [31]:
           production final clf = grid search.best estimator
In [33]:
           titanic_test_df = pd.read_csv("test.csv")
In [34]:
           final test data = pipeline.fit transform(titanic test df)
In [38]:
           final_test_data
Out[38]:
              Passengerld Pclass
                                     Age
                                         SibSp
                                                Parch
                                                         Fare
                                                                С
                                                                    S
                                                                        Q Female
                                                                                  Male
            0
                      892
                              3 34.50000
                                                        7.8292 0.0 1.0 0.0
                                                                              0.0
                                                                                    1.0
            1
                      893
                              3 47 00000
                                                    0
                                                        7 0 0 0 0 0 0 1 0
                                                                              10
                                                                                   0.0
            2
                      894
                              2 62.00000
                                             0
                                                    0
                                                        9.6875 0.0
                                                                  1.0 0.0
                                                                              0.0
                                                                                    1.0
            3
                      895
                              3 27.00000
                                                        8.6625 0.0 0.0 1.0
                                                                              0.0
                                                                                    1.0
            4
                      896
                              3 22 00000
                                             1
                                                    1
                                                       12 2875 00 00 10
                                                                              10
                                                                                    0.0
          413
                    1305
                              3 30.27259
                                                        8.0500 0.0 0.0 1.0
                                                                                    1.0
                                             0
                                                                              0.0
                              1 39.00000
          414
                     1306
                                             0
                                                    0 108.9000 1.0 0.0 0.0
                                                                              1.0
                                                                                   0.0
          415
                     1307
                              3 38.50000
                                             0
                                                        7.2500 0.0 0.0 1.0
                                                                              0.0
                                                                                    1.0
          416
                     1308
                              3 30.27259
                                                        8.0500 0.0 0.0 1.0
                                                                              0.0
                                                                                    1.0
                              3 30.27259
                                                       22.3583 1.0 0.0 0.0
          417
                    1309
                                             1
                                                    1
                                                                              0.0
                                                                                   1.0
```

418 rows × 11 columns

```
In [39]:
    final_test_data.info()
```

```
Data columns (total 11 columns):
             Column
                          Non-Null Count Dtype
              PassengerId 418 non-null
          0
                                           int64
          1
              Pclass
                           418 non-null
                                           int64
          2
                           418 non-null
                                           float64
              Aae
          3
              SibSp
                           418 non-null
                                           int64
              Parch
                           418 non-null
                                           int64
          5
                           417 non-null
                                           float64
             Fare
          6
              C
                           418 non-null
                                           float64
          7
              S
                           418 non-null
                                           float64
          8
                           418 non-null
                                           float64
              0
          9
              Female
                           418 non-null
                                           float64
          10 Male
                           418 non-null
                                           float64
         dtypes: float64(7), int64(4)
         memory usage: 36.0 KB
In [42]:
          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)
In [44]:
          predictions = production_final_clf.predict(X_data_final_test)
In [45]:
          predictions
Out[45]: array([0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0,
                1, 0, 1, 0, 1, 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, 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, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
                0, 0, 1, 1, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1,
                1, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 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, 1, 0, 1, 0, 0, 0, 1, 0, 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, 1, 1,
                                                                         1, 1, 0,
                0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 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,
                1, 0, 0, 1, 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, 1,
                0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0],
               dtype=int64)
In [46]:
          final df = pd.DataFrame(titanic test df['PassengerId'])
          final_df['Survived'] = predictions
          final_df.to_csv("predictions.csv", index = False)
In [47]:
          final df
Out[47]:
             Passengerld Survived
           0
                              0
                    892
                    893
           2
                    894
                              0
           3
                    895
                              Λ
           4
                    896
```

RangeIndex: 418 entries, 0 to 417

... 413

414

415

416

417

0

0

0

0

1305

1306

1307

1308

1309

In []:

Loading [MathJax]/jax/output/CommonHTML/fonts/TeX/fontdata.js