```
In [4]: import seaborn as sns
          import pandas as pd
          import matplotlib.pyplot as plt
          import numpy as np
 In [5]: sns.get_dataset_names()
 Out[5]: ['anagrams',
            anscombe',
            'attention',
           'brain networks',
           'car_crashes',
           'diamonds',
           'dots',
           'dowjones',
           'exercise',
           'flights',
           'fmri',
           'geyser',
            'glue',
           'ĥealthexp',
           'iris',
           'mpg',
           'penguins',
           'planets',
            'seaice',
           'taxis',
           'tips',
           'titanic']
 In [6]: data = sns.load_dataset("titanic")
 In [7]: data.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 891 entries, 0 to 890
          Data columns (total 15 columns):
           #
               Column
                             Non-Null Count
                                              Dtype
           0
               survived
                             891 non-null
                                              int64
                             891 non-null
               pclass
                                              int64
           2
                             891 non-null
                                              object
               sex
                             714 non-null
                                              float64
           3
               age
           4
               sibsp
                             891 non-null
                                              int64
               parch
                             891 non-null
                                              int64
           6
               fare
                             891 non-null
                                              float64
               embarked
                             889 non-null
                                              object
           8
               class
                             891 non-null
                                              category
           9
               who
                             891 non-null
                                              object
           10
               adult_male
                             891 non-null
                                              bool
                             203 non-null
                                              category
           11
               deck
                             889 non-null
           12
               embark_town
                                              object
           13
               alive
                             891 non-null
                                              object
                             891 non-null
              alone
                                              bool
          dtypes: bool(2), category(2), float64(2), int64(4), object(5)
          memory usage: 80.7+ KB
 In [8]: data.head()
 Out[8]:
             survived pclass
                                    age
                                        sibsp
                                               parch
                                                        fare embarked
                                                                     class
                                                                              who
                                                                                   adult_male deck
                                                                                                   embark_town
                                                                                                               alive
                                                                                                                     alone
          0
                   0
                              male
                                   22.0
                                                      7.2500
                                                                   s
                                                                      Third
                                                                                                    Southampton
                                                                                                                     False
           1
                            female
                                   38.0
                                                  0 71.2833
                                                                   С
                                                                       First
                                                                                        False
                                                                                                С
                                                                                                      Cherbourg
                                                                                                                     False
          2
                          3 female 26.0
                                            0
                                                  n
                                                     7.9250
                                                                   s
                                                                      Third
                                                                            woman
                                                                                        False
                                                                                             NaN
                                                                                                    Southampton
                                                                                                                      True
           3
                            female
                                   35.0
                                                  0 53.1000
                                                                   S
                                                                       First
                                                                            woman
                                                                                        False
                                                                                                С
                                                                                                    Southampton
                                                                                                                     False
                   n
                              male 35.0
                                            0
                                                  0 8.0500
                                                                   S
                                                                     Third
                                                                              man
                                                                                        True
                                                                                             NaN
                                                                                                    Southampton
                                                                                                                 no
                                                                                                                     True
 In [9]: len(data)
 Out[9]: 891
In [10]: | data['survived'].value_counts()
Out[10]: 0
               549
               342
          Name: survived, dtype: int64
```

```
In [11]: sns.countplot(x=data['survived'])
          plt.show()
              500
             400
           300
8
             200
             100
               0
                             ò
                                                     i
                                      survived
In [12]: sns.countplot(x=data['survived'], hue=data['pclass'])
          plt.show()
                                                           pclass
1
2
              350
             300
             250
           를 200
             150
             100
              50
                             ò
                                      survived
In [13]: sns.countplot(x=data['survived'], hue=data['sex'])
          plt.show()
                                                           sex
                                                           male
             400
             300
             200
             100
               0
                                      survived
In [14]: | sns.countplot(x=data['survived'], hue=data['embarked'])
          plt.show()
                                                         embarked
              400
                                                          ____ S
                                                          ___ c
             350
                                                          O
             300
           8 250
200
             150
             100
              50
```

survived

```
In [15]: data.columns
dtype='object')
In [16]: cols =['fare', 'class', 'who', 'adult_male', 'deck', 'embark_town', 'alive', 'alone']
In [17]: data_new = data.drop(cols, axis=1)
         data_new
Out[17]:
                                  age sibsp parch embarked
              survived pclass
                             sex
                                                       s
           0
                   0
                             male
                                 22.0
                                               0
                                               0
                                                       С
           1
                   1
                         1 female
                                 38.0
                                         1
           2
                                               0
                                 26.0
                                         0
                                                       s
                         3 female
                                               0
           3
                         1 female
                                 35.0
                                         1
                                                       s
                   0
                                         0
                                               0
                                                       s
                         3
                                 35.0
                             male
          886
                   0
                         2
                             male
                                 27.0
                                         0
                                                       s
          887
                                  19.0
                                         0
                                               0
                                                       s
                         1 female
          888
                   0
                                 NaN
                                                       s
          889
                                         0
                                               0
                                                       С
          890
                   0
                                  32.0
                                         0
                                               0
                                                       Q
         891 rows × 7 columns
In [18]: data_new.isnull().sum()
Out[18]: survived
                      0
         pclass
                      0
         sex
                      0
                    177
         age
         sibsp
                      0
         parch
                      0
         embarked
         dtype: int64
In [19]: mean_age = data_new["age"].mean()
         mean_age = np.round(mean_age, 2)
         mean_age
Out[19]: 29.7
In [20]: data_new['age'] = data_new['age'].fillna(mean_age)
In [21]: data_new.isnull().sum()
Out[21]: survived
                    0
         pclass
                    a
         sex
                    0
         age
                    0
         sibsp
         parch
                    0
         embarked
         dtype: int64
In [22]: data_new = data_new.dropna()
In [23]: data_new.isnull().sum()
Out[23]: survived
         pclass
                    0
         sex
                    0
                    0
         sibsp
         parch
         embarked
         dtype: int64
```

```
In [24]: from sklearn.preprocessing import LabelEncoder
                          enc = LabelEncoder()
                          data_new['sex'] = enc.fit_transform(data_new['sex'])
                          C:\Users\shara\AppData\Local\Temp/ipykernel_24672/67093066.py:3: 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#returning-a-view-ve
                          rsus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)
                                data_new['sex'] = enc.fit_transform(data_new['sex'])
In [25]: data_new['embarked'] = enc.fit_transform(data_new['embarked'])
                           \verb|C:\Users\shara\AppData\Local\Temp/ipykernel_24672/992371747.py: 1: SettingWithCopyWarning: Application of the property of
                          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
                          \textbf{See the caveats in the documentation:} \ \texttt{https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html\#returning-a-view-vertically and the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation in the documentation is the documentation in the documentation in the documentation is the doc
                           rsus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)
                                data_new['embarked'] = enc.fit_transform(data_new['embarked'])
In [26]: x = np.array(data_new.iloc[:,1:])
In [27]: y = np.array(data_new.iloc[:,0])
In [28]: from sklearn.model_selection import train_test_split
                           xtrain, xtest, ytrain, ytest = train_test_split(x, y, train_size = 0.8, random_state = 3)
In [29]: pd.DataFrame(y).value_counts()
Out[29]: 0
                                        549
                                        340
                          dtype: int64
In [30]: |pd.DataFrame(ytrain).value_counts()
Out[30]: 0
                                        439
                                        272
                          dtype: int64
In [31]: from sklearn.neighbors import KNeighborsClassifier
                           model = KNeighborsClassifier(n_neighbors=3, p=2)
In [32]: model.fit(xtrain, ytrain)
Out[32]: KNeighborsClassifier(n_neighbors=3)
In [33]: ypred = model.predict(xtest)
In [35]: ytest[1] == ypred[1]
Out[35]: True
In [36]: count = 0
                          for i in range(len(ytest)):
                                if ypred[i] == ytest[i]:
                                     count = count + 1
                          count
Out[36]: 133
In [37]: count/len(ytest)
Out[37]: 0.7471910112359551
In [38]: from sklearn.metrics import accuracy_score
                          a = accuracy_score(ytest,ypred)
                          а
Out[38]: 0.7471910112359551
```

```
In [39]: from sklearn.metrics import confusion_matrix
         cm = confusion_matrix(ytest, ypred)
         cm
Out[39]: array([[92, 18],
                [27, 41]], dtype=int64)
In [40]: acc = (94+45)/(94+16+23+45)
         acc
Out[40]: 0.7808988764044944
In [41]: import joblib
         joblib.dump(model, "titanic.pkl")
Out[41]: ['titanic.pkl']
 In [ ]: mymodel = joblib.load("/content/titanic.pkl")
 In [ ]: mymodel.predict(xtest)
 In [ ]: mymodel.predict([[1,1,22,2,0,1]])
 In [ ]: d={}
         for i in range(1,439):
           model = KNeighborsClassifier(n_neighbors=i, p=2)
           model.fit(xtrain, ytrain)
           ypred = model.predict(xtest)
           count = 0
           for j in range(len(ytest)):
            if ypred[j] == ytest[j]:
               count = count + 1
           d[i] = count/len(ytest)
         for i, j in d.items():
           if j == max(d.values()):
             print(i,j)
```