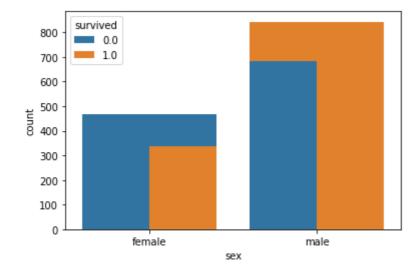
Titanic Survival Prediction

```
In [66]:
        import pandas as pd
         import matplotlib as plt
         import seaborn as sns
         import numpy as np
In [3]: dataset=pd.read csv(r"D:\Arivu\Desktop\titanic3.csv")
In [10]: dataset.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 1310 entries, 0 to 1309
         Data columns (total 14 columns):
          #
              Column
                         Non-Null Count Dtype
              -----
                         -----
              pclass
                         1309 non-null float64
          0
                         1309 non-null float64
          1
              survived
                         1309 non-null object
1309 non-null object
          2
              name
          3
              sex
          4
              age
                         1046 non-null float64
                         1309 non-null float64
          5
              sibsp
                         1309 non-null
          6
              parch
                                         float64
          7
                         1309 non-null object
              ticket
          8
              fare
                         1308 non-null float64
                         295 non-null object
1307 non-null object
          9
              cabin
          10 embarked
          11 boat
                         486 non-null object
          12 body
                         121 non-null
                                         float64
          13 home.dest 745 non-null
                                         object
         dtypes: float64(7), object(7)
         memory usage: 143.4+ KB
In [ ]:
In [4]: print(dataset.columns)
         Index(['pclass', 'survived', 'name', 'sex', 'age', 'sibsp', 'parch', 'tick
         et',
                 'fare', 'cabin', 'embarked', 'boat', 'body', 'home.dest'],
               dtype='object')
 In [6]: #Column Selection/Field Selection
         y=dataset[["survived"]]
         print(y.head())
            survived
         0
                 1.0
         1
                 1.0
         2
                 0.0
         3
                 0.0
                 0.0
         4
```

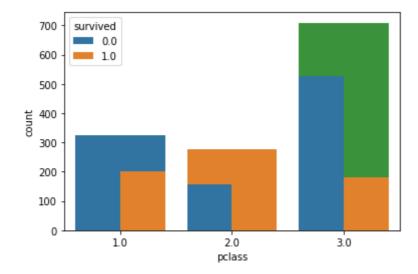
```
In [7]: gender=dataset['sex']
sns.countplot(data=dataset,x='sex')
sns.countplot(data=dataset,x='sex',hue='survived')
```

Out[7]: <AxesSubplot:xlabel='sex', ylabel='count'>



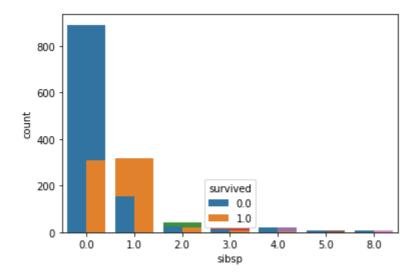
```
In [8]: sns.countplot(data=dataset,x='pclass')
sns.countplot(data=dataset,x='pclass',hue='survived')
```

Out[8]: <AxesSubplot:xlabel='pclass', ylabel='count'>



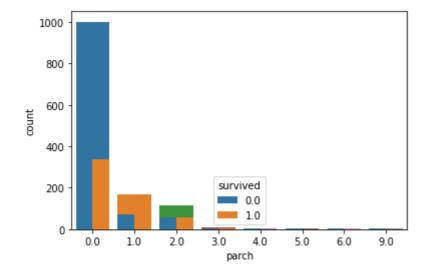
```
In [11]: sns.countplot(data=dataset,x='sibsp')
sns.countplot(data=dataset,x='sibsp',hue='survived')
```

Out[11]: <AxesSubplot:xlabel='sibsp', ylabel='count'>



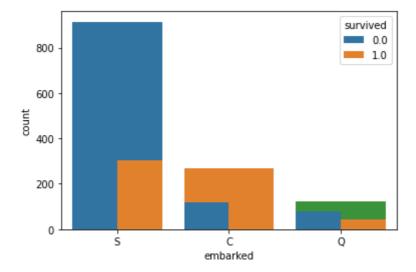
```
In [12]: sns.countplot(data=dataset,x='parch')
sns.countplot(data=dataset,x='parch',hue='survived')
```

Out[12]: <AxesSubplot:xlabel='parch', ylabel='count'>



```
In [13]: sns.countplot(data=dataset,x='embarked')
sns.countplot(data=dataset,x='embarked',hue='survived')
```

Out[13]: <AxesSubplot:xlabel='embarked', ylabel='count'>



```
In [14]: dataset.drop("name",axis=1,inplace=True) #drop Name
    dataset.drop("ticket",axis=1,inplace=True) #drop Ticket
    dataset.drop("fare",axis=1,inplace=True) #drop Fare
    dataset.drop("parch",axis=1,inplace=True) #drop parch
    dataset.drop("cabin",axis=1,inplace=True) #drop Cabin
```

```
In [18]: x=dataset[['pclass','sex','age','sibsp','embarked']]
print(x.isnull())
```

```
pclass
                   age sibsp
                             embarked
             sex
0
     False False False
                                False
1
     False False False
                                False
2
     False False False
                                False
3
     False False False
                                False
     False False False
4
                                False
                   . . .
                                  . . .
. . .
     False False
                  True False
                                False
1305
1306
     False False False
                                False
     False False False
                                False
1307
1308
     False False False
                                False
      True
1309
            True
                  True
                        True
                                 True
```

[1310 rows x 5 columns]

```
In [19]:
          #Working with Null values
          print(dataset.isnull())
          sns.heatmap(dataset.isnull())
          sns.heatmap(dataset.isnull() ,yticklabels=False , cmap="YlGnBu")
          print(dataset[dataset["age"].isnull()]) #print null value
                pclass survived
                                      sex
                                              age
                                                   sibsp
                                                          embarked
                                                                       boat
                                                                              body
                                                                                    home.
          dest
          0
                 False
                            False False
                                          False
                                                   False
                                                              False
                                                                      False
                                                                              True
                                                                                         F
          alse
          1
                 False
                            False
                                   False
                                           False
                                                   False
                                                              False
                                                                      False
                                                                              True
                                                                                         F
          alse
                 False
                                   False
                                           False
                                                                                         F
          2
                            False
                                                   False
                                                              False
                                                                       True
                                                                              True
          alse
          3
                 False
                            False False False
                                                  False
                                                              False
                                                                       True
                                                                             False
                                                                                         F
          alse
          4
                 False
                            False
                                   False False
                                                   False
                                                              False
                                                                       True
                                                                              True
                                                                                         F
          alse
          . . .
                    . . .
                                                                . . .
          . . .
          1305
                 False
                            False False
                                            True False
                                                              False
                                                                       True
                                                                              True
          True
                                          False
                                                  False
          1306
                 False
                            False False
                                                              False
                                                                       True
                                                                             False
          True
                                           False
          1307
                 False
                            False
                                   False
                                                   False
                                                              False
                                                                       True
                                                                              True
          True
          1308
                 False
                            False
                                    False
                                           False
                                                   False
                                                              False
                                                                       True
                                                                              True
          True
          1309
                  True
                             True
                                     True
                                             True
                                                    True
                                                               True
                                                                       True
                                                                              True
          True
          [1310 rows x 9 columns]
                pclass
                        survived
                                       sex
                                            age
                                                  sibsp embarked boat
                                                                         body
          15
                    1.0
                              0.0
                                      male
                                            NaN
                                                    0.0
                                                                S
                                                                   NaN
                                                                          NaN
          37
                              1.0
                                                                S
                    1.0
                                      male
                                            NaN
                                                    0.0
                                                                      9
                                                                          NaN
                                                                C
          40
                    1.0
                              0.0
                                      male
                                            NaN
                                                    0.0
                                                                   NaN
                                                                          NaN
                                                                S
          46
                    1.0
                              0.0
                                      male
                                            NaN
                                                    0.0
                                                                   NaN
                                                                          NaN
                                                                C
          59
                    1.0
                              1.0
                                   female
                                            NaN
                                                    0.0
                                                                      5
                                                                          NaN
          . . .
                    . . .
                               . . .
                                       . . .
                                                    . . .
                                                                          . . .
          1297
                    3.0
                              0.0
                                      male
                                            NaN
                                                    0.0
                                                                S
                                                                   NaN
                                                                          NaN
                              0.0
                                                                C
          1302
                    3.0
                                      male
                                            NaN
                                                    0.0
                                                                   NaN
                                                                          NaN
          1303
                    3.0
                              0.0
                                      male
                                            NaN
                                                    0.0
                                                                C
                                                                   NaN
                                                                          NaN
                    3.0
                              0.0
                                    female
                                            NaN
                                                                C
                                                                          NaN
          1305
                                                    1.0
                                                                   NaN
          1309
                              NaN
                                                              NaN
                                                                   NaN
                                                                          NaN
                    NaN
                                       NaN
                                            NaN
                                                    NaN
                        home.dest
          15
                     New York, NY
          37
                  Los Angeles, CA
          40
                Philadelphia, PA
          46
                              NaN
          59
                     New York, NY
```

[264 rows x 9 columns]

NaN

NaN

NaN

NaN

NaN

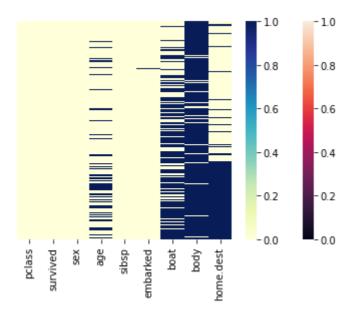
. . . 1297

1302

1303

1305

1309



```
In [20]: #here we find out the mean value using the perticular pclass
for i in range(1,4):
    age=int(dataset[dataset["pclass"]==i]['age'].dropna().mean())
    print(age)
```

39 29 24

```
In [23]: #fill the Null value
def set_age(row):
    pclass=row[0]
    age=row[1]
    if np.isnan(age):
        if pclass==1:
            return 38
        elif pclass==2:
            return 29
        else:
            return age
```

```
In [24]: dataset[['pclass', 'age']].apply(set_age,axis=1)
```

```
Out[24]: 0
                   29.0000
          1
                    0.9167
          2
                    2.0000
          3
                   30.0000
          4
                   25.0000
                    . . .
          1305
                   25.0000
          1306
                   26.5000
          1307
                   27.0000
          1308
                   29.0000
          1309
                   25.0000
          Length: 1310, dtype: float64
```

Out[25]:

```
2.0 3.0
0 0 0
```

- **1** 0 0
- **2** 0 0
- **3** 0 0
- 4 0 0

```
In [26]: sex=pd.get_dummies(dataset['sex'],drop_first=True)
    sex.head()
```

Out[26]:

	IIIaie
0	0
1	1

- 2 0
- 3 1
- **4** 0

```
In [27]: sibsp=pd.get_dummies(dataset['sibsp'],drop_first=True)
    sibsp.head()
```

Out[27]:

	1.0	2.0	3.0	4.0	5.0	8.0
0	0	0	0	0	0	0
1	1	0	0	0	0	0
2	1	0	0	0	0	0
3	1	0	0	0	0	0
4	1	0	0	0	0	0

```
In [28]: embarked=pd.get_dummies(dataset['embarked'],drop_first=True)
embarked.head()
```

Out[28]:

```
Q S 0 1
```

- **1** 0 1
- **2** 0 1
- **3** 0 1
- **4** 0 1

```
#we have to drop catagorical variable
In [30]:
         dataset.drop("pclass",axis=1,inplace=True)
         dataset.drop("sex",axis=1,inplace=True)
         dataset.drop("sibsp",axis=1,inplace=True)
         dataset.drop("embarked",axis=1,inplace=True)
In [32]: y=dataset[['survived']]
         dataset.drop("survived",axis=1,inplace=True)
In [34]:
         #use that variable that we have converted into dummy variable
         dataset=pd.concat([pclass,sex,sibsp,embarked] , axis=1)
         print(dataset.head())
             2.0
                                  2.0 3.0 4.0
                                                 5.0
                 3.0
                      male
                             1.0
                                                      8.0
                                                           Q
                                                              S
         0
              0
                   0
                         0
                               0
                                    0
                                         0
                                              0
                                                   0
                                                         0
                                                           0
                                                              1
         1
                   0
                          1
                               1
                                         0
                                              0
                                                         0
                                                              1
              0
                                    0
                                                   0
                                                           0
         2
              0
                   0
                         0
                               1
                                    0
                                         0
                                              0
                                                   0
                                                        0 0 1
         3
                               1
                                              0
              0
                   0
                          1
                                    0
                                         0
                                                   0
                                                         0
                                                           0 1
              0
                   0
                          0
                               1
                                         0
                                                         0 0 1
In [35]: #in some version of python feature name should be string
         x.columns=x.columns.astype(str)
         x.columns
Out[35]: Index(['pclass', 'sex', 'age', 'sibsp', 'embarked'], dtype='object')
In [55]: | from sklearn.linear_model import LogisticRegression
         from sklearn.model_selection import train_test_split
         x_train,x_test,y_train,y_test=train_test_split( x_encoded, y, test_size=0.2
In [56]: print(x_train)
                        age sibsp
                                     embarked sex_female
                pclass
                                                           sex_male
         1024
                   3.0 25.0
                                0.0
                                            2
                   2.0 24.0
                                1.0
                                            2
                                                                   0
         467
                                                         1
         1022
                   3.0
                       0.0
                                0.0
                                            2
                                                         0
                                                                   1
         996
                   3.0 33.0
                                0.0
                                            2
                                                         0
                                                                   1
                   3.0 15.0
                                1.0
                                            0
                                                        1
         1300
                                                                   0
                                . . .
          . . .
                   . . .
                         . . .
                                          . . .
                                                       . . .
                                                                 . . .
         1242
                   3.0
                       0.0
                                0.0
                                            0
                                                         0
                                                                   1
         924
                   3.0 34.5
                                0.0
                                            1
                                                         0
                                                                   1
         1247
                   3.0
                       0.0
                                1.0
                                            2
                                                        1
                                                                   0
         271
                  1.0
                       24.0
                                            2
                                                         0
                                1.0
                                                                   1
                                            2
         474
                   2.0 31.0
                                0.0
                                                         0
                                                                   1
         [1048 rows x 6 columns]
```

```
In [38]:
         print(y_train)
               survived
         1024
                     0.0
         467
                     1.0
         1022
                     0.0
         996
                     0.0
         1300
                     1.0
          . . .
                     . . .
         1242
                     0.0
                     0.0
         924
                     1.0
         1247
         271
                     1.0
         474
                     0.0
         [1048 rows x 1 columns]
In [39]:
         # Assuming 'Embarked' is a categorical variable
         x_encoded = pd.get_dummies(x, columns=['embarked'])
In [42]: from sklearn.preprocessing import LabelEncoder
         label_encoder = LabelEncoder()
         x['embarked'] = label_encoder.fit_transform(x['embarked'])
         C:\Users\arivu\AppData\Local\Temp\ipykernel_5356\1609296811.py:4: SettingW
         ithCopyWarning:
         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-doc
         s/stable/user guide/indexing.html#returning-a-view-versus-a-copy (https://
         pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
         view-versus-a-copy)
           x['embarked'] = label_encoder.fit_transform(x['embarked'])
In [45]: | # Assuming 'Sex' is a categorical variable
         x_encoded = pd.get_dummies(x, columns=['sex'])
```

```
In [46]: from sklearn.preprocessing import LabelEncoder
         label_encoder = LabelEncoder()
         x['sex'] = label_encoder.fit_transform(x['sex'])
         C:\Users\arivu\AppData\Local\Temp\ipykernel_5356\1486867810.py:4: SettingW
         ithCopyWarning:
         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-doc
         s/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://
         pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
         view-versus-a-copy)
           x['sex'] = label_encoder.fit_transform(x['sex'])
In [58]: x_encoded.fillna(0, inplace=True) # Replace NaN values with 0
         y_train.fillna(0, inplace=True)
In [59]: print(x_encoded.isnull().sum())
         print(y_train.isnull().sum())
         pclass
                       0
         age
                       0
         sibsp
                       0
         embarked
                       0
         sex female
                       0
         sex_male
         dtype: int64
         survived
         dtype: int64
In [49]: model=LogisticRegression()
In [60]: |model.fit(x_train,y_train)
         C:\Users\arivu\AppData\Local\Programs\Python\Python310\lib\site-packages\s
         klearn\utils\validation.py:993: DataConversionWarning: A column-vector y w
         as passed when a 1d array was expected. Please change the shape of y to (n
         samples, ), for example using ravel().
           y = column_or_1d(y, warn=True)
Out[60]: LogisticRegression()
In [61]: print(model.coef )
         [[-0.89328966 -0.01041016 -0.24473532 -0.26075279 1.69867438 -0.9113117
         6]]
In [63]: y_pred=model.predict(x_test)
```

```
print(y_pred)
In [64]:
         [1. 0. 1. 0. 0. 0. 1. 0. 0. 0. 0. 1. 1. 1. 1. 1. 0. 0. 0. 1. 1. 0. 0. 1.
          0. 0. 1. 1. 0. 0. 0. 0. 0. 0. 1. 0. 0. 1. 0. 0. 1. 1. 1. 1. 0. 1. 1.
          0. 1. 0. 0. 0. 1. 0. 1. 0. 0. 0. 0. 0. 1. 1. 1. 0. 0. 0. 0. 1. 0.
          0. 1. 0. 1. 0. 0. 1. 0. 0. 0. 0. 0. 1. 1. 0. 1. 1. 0. 0. 1. 1. 1. 0.
          0. 0. 0. 0. 1. 0. 1. 0. 0. 0. 0. 1. 1. 0. 0. 0. 1. 0. 0. 0. 0. 1. 1. 1.
          0. 0. 0. 0. 0. 1. 1. 0. 1. 0. 0. 1. 0. 1. 0. 1. 0. 1. 0. 0. 1. 1.
          0. 1. 0. 0. 0. 0. 0. 0. 1. 1. 1. 0. 0. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0.
          0. 0. 0. 1. 0. 0. 0. 1. 0. 0. 1. 0. 1. 1. 1. 0. 0. 1. 0. 0. 0. 0. 0. 0.
          0. 0. 0. 0. 1. 1. 1. 0. 0. 1. 0. 0. 1. 0. 1. 0. 1. 0. 1. 1. 0. 0. 1.
          1. 1. 1. 0. 0. 0. 1. 0. 0. 0. 0. 1. 0. 1. 0. 0. 0. 0. 0. 0. 0. 1. 0. 0. 1.
          0. 0. 0. 1. 1. 0. 0. 1. 1. 0. 0. 0. 1. 1. 1. 0. 0. 0. 0. 0. 1. 0.]
In [65]: #confusion matrix
         from sklearn.metrics import confusion_matrix
         print(confusion_matrix(y_test,y_pred))
         [[131 21]
          [ 39 71]]
 In [ ]:
 In [ ]:
```