```
In [1]: # Welcome to my first submission on Kaggle
```

### **TITANIC DATA ANALYSIS**

import math

```
In [2]: # Collection of Data
## we start by importing all necessary modulles

In [3]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

In [4]: train = pd.read\_csv('train.csv')
 train.head(5)

#### Out[4]:

	traveller_ID	ticket_class	name	sex	age	Siblings_s <sub>I</sub>	oouses	Parchil	TickNum	1
0	1214	3	Smiljanic, Mr. Mile	male	NaN		0	0	315037	8.6
1	677	3	Bostandyeff, Mr. Guentcho	male	26.0		0	0	349224	7.8
2	534	2	Phillips, Miss. Kate Florence ("Mrs Kate Louis	female	19.0		0	0	250655	26.0
3	1174	3	Sage, Miss. Dorothy Edith "Dolly"	female	NaN		8	2	CA. 2343	69.5
4	864	3	Henriksson, Miss. Jenny Lovisa	female	28.0		0	0	347086	7.7
4										<b>&gt;</b>

```
In [5]: test = pd.read_csv('test.csv')
  test.head()
```

#### Out[5]:

	traveller_ID	ticket_class	name	sex	age	Siblings_spouses	Parchil	TickNum	f
0	1148	3.0	Rintamaki, Mr. Matti	male	35.0	0.0	0.0	STON/O 2. 3101273	7.1;
1	1049	3.0	Nakid, Mr. Sahid	male	20.0	1.0	1.0	2653	15.74
2	982	3.0	Lyntakoff, Mr. Stanko	male	NaN	0.0	0.0	349235	7.8!
3	808	3.0	Ford, Mr. Arthur	male	NaN	0.0	0.0	A/5 1478	8.0
4	1195	3.0	Shaughnessy, Mr. Patrick	male	NaN	0.0	0.0	370374	7.7
4									•

In [6]: sample\_submission = pd.read\_csv('sample\_submission.csv')
 sample\_submission.head(5)

#### Out[6]:

		traveller_ID	Survived
	0	1148	0
	1	1049	1
	2	982	0
;	3	808	1
	4	1195	0

```
In [7]: ## To get total number of passengers on board
print('Number of passengers on board is, ' + str(len(train.index)))
```

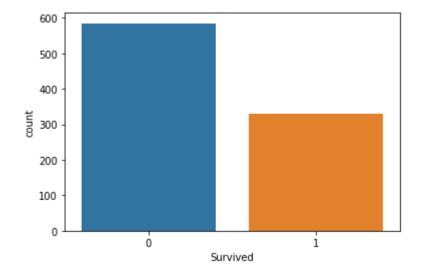
Number of passengers on board is, 916

# **Analysing of Data**

```
In [8]: # number of those who survived
```

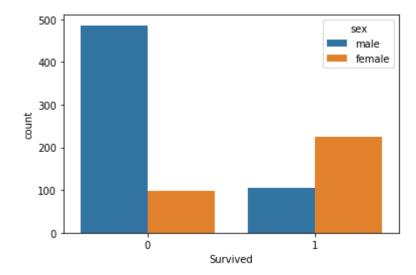
```
In [9]: sns.countplot( x="Survived", data=train)
```

Out[9]: <matplotlib.axes.\_subplots.AxesSubplot at 0xadfae10>



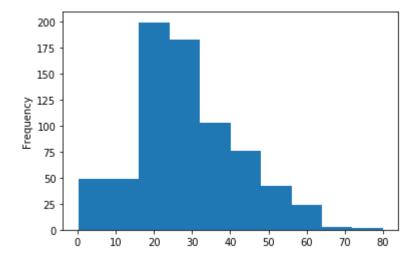
In [10]: # number of survivals who were males, and females
 sns.countplot(x="Survived", hue="sex", data=train)

Out[10]: <matplotlib.axes.\_subplots.AxesSubplot at 0xaebae48>



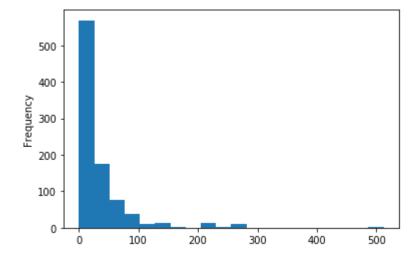
```
In [11]: #visualise age distribution
    train["age"].plot.hist()
```

Out[11]: <matplotlib.axes.\_subplots.AxesSubplot at 0xaf4a4a8>



In [12]: # Visualise fare of travellers
 train["fare"].plot.hist(bins=20, )

Out[12]: <matplotlib.axes.\_subplots.AxesSubplot at 0xaff60f0>



```
In [13]: #To obtain remaining information
         train.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 916 entries, 0 to 915
         Data columns (total 13 columns):
                             916 non-null int64
         traveller ID
         ticket_class
                              916 non-null int64
                              916 non-null object
         name
                              916 non-null object
         sex
         age
                              729 non-null float64
                              916 non-null int64
         Siblings_spouses
         Parchil
                              916 non-null int64
         TickNum
                              916 non-null object
         fare
                              915 non-null float64
                              204 non-null object
         cabin
         embarked
                              915 non-null object
         MedBoat
                              327 non-null object
                              916 non-null int64
         Survived
         dtypes: float64(2), int64(5), object(6)
         memory usage: 93.1+ KB
In [14]: train.shape, test.shape
Out[14]: ((916, 13), (393, 12))
In [15]: train.describe()
Out[15]:
```

	traveller_ID	ticket_class	age	Siblings_spouses	Parchil	fare	Survi
count	916.000000	916.000000	729.000000	916.000000	916.000000	915.000000	916.0000
mean	656.248908	2.305677	29.102309	0.539301	0.386463	33.709221	0.3610
std	381.176191	0.841811	13.866954	1.082188	0.893933	52.840656	0.480€
min	0.000000	1.000000	0.166700	0.000000	0.000000	0.000000	0.0000
25%	328.250000	2.000000	21.000000	0.000000	0.000000	7.925000	0.0000
50%	670.000000	3.000000	28.000000	0.000000	0.000000	14.500000	0.0000
75%	974.250000	3.000000	37.000000	1.000000	0.000000	31.275000	1.0000
max	1308.000000	3.000000	80.000000	8.000000	9.000000	512.329200	1.0000
4							<b>•</b>

## **DATA CLEANING**

In [16]: #we proceed to cleaning of data by removal of categorical datas and replacing of missing values

In [17]: train.isnull().head()

Out[17]:

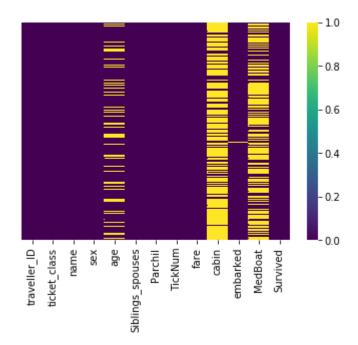
	traveller_ID	ticket_class	name	sex	age	Siblings_spouses	Parchil	TickNum	fare	cab
0	False	False	False	False	True	False	False	False	False	Tr
1	False	False	False	False	False	False	False	False	False	Tr
2	False	False	False	False	False	False	False	False	False	Tr
3	False	False	False	False	True	False	False	False	False	Tr
4	False	False	False	False	False	False	False	False	False	Tr
4										•

In [18]: | train.isnull().sum()

Out[18]: traveller\_ID 0 ticket\_class 0 name 0 sex 0 187 age Siblings\_spouses 0 Parchil 0 TickNum 0 fare 1 cabin 712 embarked 1 MedBoat 589 Survived 0 dtype: int64

In [19]: sns.heatmap(train.isnull(), yticklabels = False , cmap='viridis' )

Out[19]: <matplotlib.axes.\_subplots.AxesSubplot at 0xb0c7828>



```
In [20]: def impute age(cols):
             age = cols[0]
             ticket class = cols[1]
             if pd.isnull(age):
                  if ticket class == 1:
                      return 37
                 elif ticket_class == 2:
                      return 28
                 else:
                      return 21
             else:
                  return age
         train['age'] = train[['age', 'ticket_class']].apply(impute_age,axis=1)
In [21]:
In [22]:
         train.embarked.fillna('S', inplace=True)
In [23]:
         train.cabin.fillna(0, inplace=True)
         train.fare.fillna(0, inplace=True)
In [24]:
In [25]:
         train.MedBoat.fillna(0, inplace=True)
In [26]: | train.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 916 entries, 0 to 915
         Data columns (total 13 columns):
         traveller ID
                              916 non-null int64
         ticket class
                              916 non-null int64
                              916 non-null object
         name
         sex
                              916 non-null object
                              916 non-null float64
         age
                              916 non-null int64
         Siblings_spouses
         Parchil
                              916 non-null int64
                              916 non-null object
         TickNum
                              916 non-null float64
         fare
         cabin
                              916 non-null object
         embarked
                              916 non-null object
         MedBoat
                              916 non-null object
         Survived
                              916 non-null int64
         dtypes: float64(2), int64(5), object(6)
         memory usage: 93.1+ KB
```

In [27]: train['cabin'].value\_counts()

Out[27]:	0 B57 B59 G6 C23 C25 C22 C26 C101 B58 B60 F2 F33 A34 E33 C89 D D21 C92 C65 B5 C6 C54 D17 D33 D35 C85 E50 B49 B20	B63	B66	712 5 4 4 3 3 3 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
	E46 C123 C83 D36			2 2 2 2 2
	B94 C45 B10 B38 C86 D22 E52 C106 D37 B61 E34 E38 B101 D46 A26 B79 A23 F E69 B41 C110 C126 D47 E36 A7 C31 C97			

F G73 1 E10 1 E60 1 C99 1

Name: cabin, Length: 147, dtype: int64

In [28]: train['cabin'].value\_counts()

Out[28]:	0 B57 B59 G6 C23 C25 C22 C26 C101 B58 B60 F2 F33 A34 E33 C89 D D21 C92 C65 B5 C6 C54 D17 D33 D35 C85 E50 B49 B20 E46 C123 C83 D36	B63	B66	712 5 4 4 3 3 3 3 2 2 2 2 2 2 2 2 2 2 2 2 2 2
	B94 C45 B10 B38 C86 D22 E52 C106 D37 B61 E34 E38 B101 D46 A26 B79 A23 F E69 B41 C110 C126 D47 E36 A7 C31 C97			1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1

F G73 1 E10 1 E60 1 C99 1

Name: cabin, Length: 147, dtype: int64

In [29]: train['embarked'].value\_counts()

Out[29]: S 639 C 187

Q 90

Name: embarked, dtype: int64

In [30]: train['TickNum'].value\_counts()

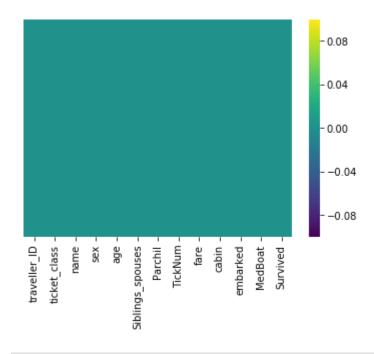
Out[30]:	CA. 2343 PC 17608 347077 3101295 CA 2144 W./C. 6608 113781 349909 S.O.C. 14879 4133 1601 347082 382652 C.A. 33112 19950 24160 347088 2650 371110 2666 W./C. 6607 26360 220845 347080 C.A. 37671 PC 17572 PC 17757 110152 2668 12749	97665555555444333333333333333333333333333
	7598 315097 350054 345770 350050 315085 349243 2691 7540 112379 237668 384461 7548 SOTON/OQ 3101316 17764 335677 368573 237735 315083 S.C./A.4. 23567 349910 233639 113807 2223 28133 SCO/W 1585	

S.P. 3464 1 SC/AH Basle 541 1 244252 1 370377 1

Name: TickNum, Length: 694, dtype: int64

In [31]: sns.heatmap(train.isnull(), yticklabels = False , cmap='viridis' )

Out[31]: <matplotlib.axes.\_subplots.AxesSubplot at 0xb3dc550>



In [32]: train.head()

#### Out[32]:

	traveller_ID	ticket_class	name	sex	age	Siblings_s <sub> </sub>	oouses	Parchil	TickNum	f
0	1214	3	Smiljanic, Mr. Mile	male	21.0		0	0	315037	8.6
1	677	3	Bostandyeff, Mr. Guentcho	male	26.0		0	0	349224	7.89
2	534	2	Phillips, Miss. Kate Florence ("Mrs Kate Louis	female	19.0		0	0	250655	26.00
3	1174	3	Sage, Miss. Dorothy Edith "Dolly"	female	21.0		8	2	CA. 2343	69.5
4	864	3	Henriksson, Miss. Jenny Lovisa	female	28.0		0	0	347086	7.7
4										•

```
In [33]: train['embarked'].value_counts()
Out[33]: S
              639
              187
         C
         Q
               90
         Name: embarked, dtype: int64
In [34]: | train['sex'].value_counts()
Out[34]: male
                   592
         female
                   324
         Name: sex, dtype: int64
In [35]: Sex = pd.get_dummies(train['sex'], drop_first=True)
         Sex.head()
Out[35]:
             male
          0
               1
          1
               1
          2
               0
               0
               0
In [36]: Embarked = pd.get_dummies(train['embarked'], drop_first=True)
         Embarked.head()
Out[36]:
             Q S
            0 1
            0 1
            0 1
            0 1
In [37]: Ticket_class = pd.get_dummies(train['ticket_class'], drop_first=True)
         Ticket_class.head()
Out[37]:
             2 3
          0 0 1
          1 0 1
          2 1 0
          3 0 1
          4 0 1
```

In [38]: train.head()

Out[38]:

	traveller_ID	ticket_class	name	sex	age	Siblings_sp	ouses	Parchil	TickNum	f
0	1214	3	Smiljanic, Mr. Mile	male	21.0		0	0	315037	8.60
1	677	3	Bostandyeff, Mr. Guentcho	male	26.0		0	0	349224	7.89
2	534	2	Phillips, Miss. Kate Florence ("Mrs Kate Louis	female	19.0		0	0	250655	26.00
3	1174	3	Sage, Miss. Dorothy Edith "Dolly"	female	21.0		8	2	CA. 2343	69.5
4	864	3	Henriksson, Miss. Jenny Lovisa	female	28.0		0	0	347086	7.7
										•

In [39]: train\_drop = train.drop(['name','traveller\_ID', 'ticket\_class', 'sex', 'TickNu
m', 'cabin','embarked',], axis=1, inplace=True)

In [40]: train.head()

Out[40]:

	age	Siblings_spouses	Parchil	fare	MedBoat	Survived
0	21.0	0	0	8.6625	0	0
1	26.0	0	0	7.8958	0	0
2	19.0	0	0	26.0000	11	1
3	21.0	8	2	69.5500	0	0
4	28.0	0	0	7.7750	0	0

In [41]: train\_new = pd.concat([train, Sex, Embarked, Ticket\_class, ], axis=1, )

In [42]: train\_new.head()

Out[42]:

	age	Siblings_spouses	Parchil	fare	MedBoat	Survived	male	Q	S	2	3	
0	21.0	0	0	8.6625	0	0	1	0	1	0	1	
1	26.0	0	0	7.8958	0	0	1	0	1	0	1	
2	19.0	0	0	26.0000	11	1	0	0	1	1	0	
3	21.0	8	2	69.5500	0	0	0	0	1	0	1	
4	28.0	0	0	7.7750	0	0	0	0	1	0	1	

## **Testing and Trainning of Data**

```
In [43]: | X = train_new.drop(['Survived'], axis=1 )
          y = train_new['Survived']
In [44]: | # for i in X['MedBoat']:
                if type(i) == 'str':
          X['MedBoat'].replace(['A','B','C','D'], 0, inplace=True)
          X['MedBoat'].replace(['13 15', '8 10', '15 16', '5 7'], [14, 9, 15, 6], inplac
          e=True)
          # X['MedBoat'] = X['MedBoat'].map({})
In [45]: X['MedBoat'].value_counts()
Out[45]: 0
                642
                 30
         13
          15
                 25
          4
                 22
          10
                 21
          11
                 20
          5
                 20
          3
                 18
         14
                 17
          9
                 16
          16
                 14
          7
                 13
          8
                 13
          6
                 12
          2
                 12
         12
                 11
                  5
          1
                  2
          14
          15
                  1
                  1
         Name: MedBoat, dtype: int64
In [46]: y.head()
Out[46]: 0
               0
          1
               0
               1
          3
         Name: Survived, dtype: int64
In [47]:
         SEED = 1000
In [48]: from sklearn.model_selection import train_test_split
```

```
In [49]: X_train, X_test, y_train, y_test = train_test_split(X,
                                                              y, test_size=0.30,
                                                              random state=SEED)
In [50]: from catboost import CatBoostClassifier
In [51]:
         X train.head()
Out[51]:
                   Siblings_spouses Parchil
                                            fare MedBoat male Q S 2 3
               age
          520 24.0
                                       0 82.2667
                                                       7
                                                               0 1 0 0
                                1
          198 21.0
                                0
                                          8.1125
                                                      13
                                                               0 1 0 1
          554 25.0
                                1
                                       0 17.8000
                                                              0 1 0 1
                                                       0
          300 42.0
                                0
                                       0 13.0000
                                                       0
                                                              0 1 1 0
          863 31.0
                                       0 57.0000
                                                       3
                                                            1 0 1 0 0
In [52]:
         %%time
         params = {'loss_function':'Logloss', # objective function
                    'eval metric':'AUC', # metric
                    'verbose': 1000, # output to stdout info about training process ever
         y 200 iterations
                    'random seed': SEED
                  }
         cbc 1 = CatBoostClassifier(**params)
         cbc 1.fit(X train, y train, # data to train on (required parameters, unless we
         provide X as a pool object, will be shown below)
                    eval_set=(X_test, y_test), # data to validate on
                   use best model=True, # True if we don't want to save trees created a
         fter iteration with the best validation score
                   plot=True # True for visualization of the training process (it is no
         t shown in a published kernel - try executing this code)
                  );
         Learning rate set to 0.028392
         0:
                 test: 0.9614032 best: 0.9614032 (0)
                                                          total: 49.6ms
                                                                          remaining: 4
         9.5s
         999:
                 test: 0.9758446 best: 0.9779488 (56)
                                                          total: 2.44s
                                                                          remaining: 0u
         s
         bestTest = 0.9779488066
         bestIteration = 56
         Shrink model to first 57 iterations.
         Wall time: 3.06 s
Out[52]: <catboost.core.CatBoostClassifier at 0xae03400>
In [53]: predictions = cbc 1.predict(X test)
```

```
In [54]:
          from sklearn.metrics import classification report, confusion matrix
In [55]:
          print(classification_report(y_test,predictions))
                          precision
                                         recall f1-score
                                                              support
                       0
                                0.95
                                           0.99
                                                      0.97
                                                                   177
                       1
                                0.99
                                           0.90
                                                                    98
                                                      0.94
                                                      0.96
                                                                   275
               accuracy
                                0.97
                                           0.95
                                                      0.96
                                                                   275
              macro avg
                                0.96
                                           0.96
                                                      0.96
                                                                   275
          weighted avg
In [56]:
          test.head()
Out[56]:
              traveller_ID ticket_class
                                            name
                                                         age
                                                              Siblings_spouses Parchil TickNum
                                                   sex
                                                                                       STON/O
                                        Rintamaki,
           0
                    1148
                                 3.0
                                                        35.0
                                                                          0.0
                                                                                  0.0
                                                   male
                                                                                                 7.12
                                                                                            2.
                                          Mr. Matti
                                                                                       3101273
                                         Nakid, Mr.
                    1049
                                 3.0
                                                  male
                                                        20.0
                                                                          1.0
                                                                                  1.0
                                                                                          2653
                                                                                               15.74
                                            Sahid
                                      Lyntakoff, Mr.
                     982
                                 3.0
           2
                                                   male
                                                        NaN
                                                                          0.0
                                                                                  0.0
                                                                                        349235
                                                                                                 7.89
                                           Stanko
                                          Ford, Mr.
                     808
           3
                                 3.0
                                                   male NaN
                                                                          0.0
                                                                                  0.0
                                                                                       A/5 1478
                                                                                                 8.0
                                            Arthur
                                     Shaughnessy,
                    1195
                                 3.0
                                                   male
                                                        NaN
                                                                          0.0
                                                                                  0.0
                                                                                        370374
                                                                                                 7.7
                                        Mr. Patrick
In [57]:
          def impute_age(cols):
               age = cols[0]
               ticket class = cols[1]
               if pd.isnull(age):
                   if ticket class == 1:
                        return 37
                   elif ticket class == 2:
                        return 28
                   else:
                        return 21
               else:
                    return age
          test['age'] = test[['age', 'ticket_class']].apply(impute_age,axis=1)
In [58]:
In [59]:
          test.embarked.fillna('S', inplace=True)
```

```
In [60]: test.fare.fillna(0, inplace=True)
         test.MedBoat.fillna(0, inplace=True)
In [61]:
In [62]: # for i in X['MedBoat']:
                if type(i) == 'str':
         test['MedBoat'].replace(['A','B','C','D', 'C D'], 0, inplace=True)
         test['MedBoat'].replace(['13 15 B', '5 7', '5 9'], [1, 1, 1, ], inplace=True)
         \# X['MedBoat'] = X['MedBoat'].map({})
In [63]: test['MedBoat'].value counts()
Out[63]: 0
                261
         14
                 16
         15
                 12
                 10
         7
         8
                 10
         4
                 9
                 9
         16
                 9
         9
                  9
         13
         12
                  8
                  8
         6
         3
                  8
                  8
         10
                  7
         5
         11
                  5
                  3
         1
         2
                  1
         Name: MedBoat, dtype: int64
In [64]: | test.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 393 entries, 0 to 392
         Data columns (total 12 columns):
         traveller ID
                              393 non-null int64
         ticket_class
                              393 non-null float64
         name
                              393 non-null object
                              393 non-null object
         sex
                              393 non-null float64
         age
         Siblings spouses
                              393 non-null float64
                              393 non-null float64
         Parchil
         TickNum
                              393 non-null object
                              393 non-null float64
         fare
         cabin
                              91 non-null object
         embarked
                              393 non-null object
         MedBoat
                              393 non-null object
         dtypes: float64(5), int64(1), object(6)
         memory usage: 36.9+ KB
```

```
In [65]: Ticket_class2 = pd.get_dummies(test['ticket_class'], drop_first=True)
          Ticket class2.head()
Out[65]:
             2.0 3.0
          0
              0
                  1
              0
                  1
          2
              0
                  1
          3
              0
                  1
              0
                  1
         Sex2 = pd.get_dummies(test['sex'], drop_first=True)
In [66]:
          Sex2.head()
Out[66]:
             male
          0
                1
          1
                1
          2
                1
          3
                1
                1
         Embarked2 = pd.get_dummies(test['embarked'], drop_first=True)
In [67]:
          Embarked2.head()
Out[67]:
             Q S
             0 1
          2 0 1
          3 0 1
            1 0
In [68]: | test_drop = test.drop(['name','traveller_ID', 'ticket_class', 'sex', 'TickNum'
          , 'cabin','embarked',], axis=1, inplace=True)
In [69]: | test_drop
In [70]: test_new = pd.concat([test, Sex2, Embarked2, Ticket_class2, ], axis=1, )
```

In [71]: test\_new.head(50)

### Out[71]:

	age	Siblings_spouses	Parchil	fare	MedBoat	male	Q	s	2.0	3.0
0	35.0000	0.0	0.0	7.1250	0	1	0	1	0	1
1	20.0000	1.0	1.0	15.7417	0	1	0	0	0	1
2	21.0000	0.0	0.0	7.8958	0	1	0	1	0	1
3	21.0000	0.0	0.0	8.0500	0	1	0	1	0	1
4	21.0000	0.0	0.0	7.7500	0	1	1	0	0	1
5	45.0000	0.0	0.0	26.5500	9	1	0	1	0	0
6	25.0000	0.0	0.0	7.9250	0	1	0	1	0	1
7	31.0000	0.0	0.0	13.0000	9	1	0	1	1	0
8	34.5000	0.0	0.0	7.8292	0	1	1	0	0	1
9	33.0000	1.0	0.0	53.1000	5	0	0	1	0	0
10	21.0000	8.0	2.0	69.5500	0	1	0	1	0	1
11	21.0000	2.0	0.0	21.6792	0	1	0	0	0	1
12	14.0000	1.0	0.0	11.2417	0	0	0	0	0	1
13	21.0000	0.0	0.0	7.8958	0	1	0	1	0	1
14	33.0000	0.0	0.0	7.8958	0	1	0	1	0	1
15	21.0000	0.0	0.0	7.7500	0	0	1	0	0	1
16	21.0000	0.0	0.0	7.2292	0	1	0	0	0	1
17	26.0000	0.0	0.0	7.8542	13	0	0	1	0	1
18	55.0000	0.0	0.0	135.6333	8	0	0	0	0	0
19	26.0000	1.0	2.0	20.5750	0	1	0	1	0	1
20	22.0000	0.0	0.0	9.8375	0	0	0	1	0	1
21	33.0000	0.0	0.0	5.0000	0	1	0	1	0	0
22	21.0000	0.0	0.0	7.8542	0	1	0	1	0	1
23	35.0000	0.0	0.0	26.2875	5	1	0	1	0	0
24	22.0000	0.0	0.0	7.2250	6	1	0	0	0	1
25	9.0000	1.0	1.0	15.2458	0	0	0	0	0	1
26	44.0000	0.0	0.0	13.0000	0	1	0	1	1	0
27	45.0000	0.0	2.0	30.0000	12	0	0	1	1	0
28	24.0000	0.0	0.0	7.7500	15	0	1	0	0	1
29	58.0000	0.0	0.0	146.5208	0	0	0	0	0	0
30	30.0000	0.0	0.0	12.7375	14	1	0	0	1	0
31	0.6667	1.0	1.0	14.5000	4	1	0	1	1	0
32	21.0000	2.0	0.0	23.2500	16	0	1	0	0	1
33	47.0000	1.0	0.0	61.1750	4	0	0	1	0	0
34	17.0000	0.0	0.0	73.5000	0	1	0	1	1	0

Siblings\_spouses Parchil

```
28.0000
                                                   9.5000
                                                                 0
                                                                                  0
            35
                                     0.0
                                             0.0
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                                                                           0
                                                                              1
                                                                                       1
            36
               37.0000
                                     1.0
                                             0.0
                                                  90.0000
                                                                 14
                                                                              0
                                                                                  0
                                                                                       0
                                                                        0
                                                                           1
               36.0000
                                     0.0
                                             0.0
                                                  13.0000
                                                                                       0
                                                                        1
                                                                           0
                                                                                  1
               39.0000
                                     1.0
                                             1.0
                                                  83.1583
                                                                 14
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                                                                              0
                                                                                  0
                                                                                       0
            38
                                                                        0
              37.0000
                                     0.0
                                             0.0
                                                                 0
            39
                                                  27.7208
                                                                           0
                                                                              0
                                                                                  0
                                                                                       0
               32.0000
                                     0.0
                                             0.0
                                                  13.0000
                                                                  9
            40
                                                                        0
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                                                                                  1
                                                                                       0
            41
               19.0000
                                     0.0
                                             0.0
                                                   8.0500
                                                                 0
                                                                           0
                                                                              1
                                                                                  0
                                                                                       1
                                                                        1
               42.0000
                                     0.0
                                             0.0
                                                  13.0000
                                                                 0
                                                                                  1
            43
               26.0000
                                     2.0
                                             0.0
                                                   8.6625
                                                                 0
                                                                                  0
                                                                        1
                                                                           0
                                                                                       1
               16.0000
                                     0.0
                                             0.0
                                                   9.2167
                                                                  0
            44
                                                                        1
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                                                                                       1
               21.0000
                                     0.0
                                             0.0
                                                   7.5500
                                                                  0
            45
                                                                        1
                                                                           0
                                                                                  0
                                                                                       1
            46
              32.5000
                                     0.0
                                             0.0
                                                   9.5000
                                                                 0
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                                                                                       1
               18.0000
                                     0.0
                                             0.0
                                                   8.0500
                                                                           0
                                                                                  0
                                                                                       1
            48
               27.0000
                                     0.0
                                             0.0
                                                   10.5000
                                                                 16
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                                                                              1
                                                                                  1
                                                                                       0
                                                                        O
            49 22.0000
                                     0.0
                                             0.0
                                                   7.7500
                                                                 13
                                                                          0
                                                                             1
                                                                                  0
                                                                                       1
                                                                        0
In [72]: test_new.info()
           <class 'pandas.core.frame.DataFrame'>
           RangeIndex: 393 entries, 0 to 392
           Data columns (total 10 columns):
           age
                                  393 non-null float64
           Siblings_spouses
                                  393 non-null float64
           Parchil
                                  393 non-null float64
           fare
                                  393 non-null float64
           MedBoat
                                  393 non-null object
           male
                                  393 non-null uint8
           Q
                                  393 non-null uint8
           S
                                  393 non-null uint8
           2.0
                                  393 non-null uint8
           3.0
                                  393 non-null uint8
           dtypes: float64(4), object(1), uint8(5)
           memory usage: 17.3+ KB
In [76]: | test_new.columns
Out[76]: Index([
                                                                            'Parchil',
                                  'age', 'Siblings_spouses',
                                 'fare',
                                                     'MedBoat',
                                                                                'male',
                                                                                   2.0,
                                                            'S',
                                    'Q',
                                    3.0],
                 dtype='object')
           test new.columns = ['age', 'Siblings spouses', 'Parchil', 'fare', 'male', 'MedB
           oat', 'Q', 'S', 2,3]
```

fare MedBoat male Q S 2.0 3.0

```
In [78]: test_new.head()
```

#### Out[78]:

	age	Siblings_spouses	Parchil	fare	male	MedBoat	Q	S	2	3
0	35.0	0.0	0.0	7.1250	0	1	0	1	0	1
1	20.0	1.0	1.0	15.7417	0	1	0	0	0	1
2	21.0	0.0	0.0	7.8958	0	1	0	1	0	1
3	21.0	0.0	0.0	8.0500	0	1	0	1	0	1
4	21.0	0.0	0.0	7.7500	0	1	1	0	0	1

```
In [79]: preditcs = cbc_1.predict(test_new)
```

```
In [80]: ts = pd.read_csv('test.csv')
    ts.head()
```

#### Out[80]:

	traveller_ID	ticket_class	name	sex	age	Siblings_spouses	Parchil	TickNum	f
(	<b>)</b> 1148	3.0	Rintamaki, Mr. Matti	male	35.0	0.0	0.0	STON/O 2. 3101273	7.1;
•	<b>1</b> 1049	3.0	Nakid, Mr. Sahid	male	20.0	1.0	1.0	2653	15.74
:	982	3.0	Lyntakoff, Mr. Stanko	male	NaN	0.0	0.0	349235	7.8!
;	808	3.0	Ford, Mr. Arthur	male	NaN	0.0	0.0	A/5 1478	8.0
4	<b>1</b> 1195	3.0	Shaughnessy, Mr. Patrick	male	NaN	0.0	0.0	370374	7.7!
4									•

```
In [81]: predictDF = pd.DataFrame(preditcs, columns=['Survived'])
    submit = pd.concat([ts['traveller_ID'], predictDF], axis=1)
    submit.head()
    submit.to_csv('Fourth_ksubmission.csv', index=False)
```