```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

from warnings import filterwarnings
filterwarnings(action="ignore")

#loading dataset

pd.set_option('display.max_columns',10,"display.width",1000)
train=pd.read_csv("train.csv")
test=pd.read_csv("test.csv")
train.head()
```

	PassengerId	Survived	Pclass	Name	Sex	 Parch	Ticket	Fare	Cat
0	1	0	3	Braund, Mr. Owen Harris	male	 0	A/5 21171	7.2500	N
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	 0	PC 17599	71.2833	С
4									•

```
train.shape
```

(891, 12)

test.shape

(418, 11)

train.isnull().sum()

PassengerId 0 Survived 0 Pclass 0 Name 0 Sex 0 Age 177 SibSp Parch 0 Ticket 0 0 Fare Cabin 687 Embarked 2 dtype: int64

test.isnull().sum()

PassengerId Pclass Name Sex 0 86 Age 0 SibSp Parch Ticket 0 0 Fare 1 Cabin 327 Embarked dtype: int64

#descrription of dataset
train.describe(include='all')

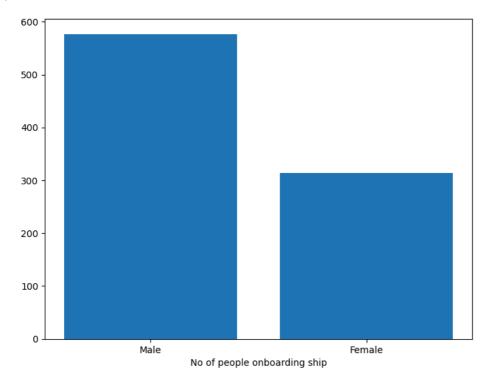
	PassengerId	Survived	Pclass	Name	Sex	•••	Parch	Ticket	
count	891.000000	891.000000	891.000000	891	891		891.000000	891	89
unique	NaN	NaN	NaN	891	2		NaN	681	
top	NaN	NaN	NaN	Braund, Mr. Owen Harris	male		NaN	347082	
freq	NaN	NaN	NaN	1	577		NaN	7	
mean	446.000000	0.383838	2.308642	NaN	NaN		0.381594	NaN	3:
std	257.353842	0.486592	0.836071	NaN	NaN		0.806057	NaN	4!
min	1.000000	0.000000	1.000000	NaN	NaN		0.000000	NaN	(
25%	223.500000	0.000000	2.000000	NaN	NaN		0.000000	NaN	;
50%	446.000000	0.000000	3.000000	NaN	NaN		0.000000	NaN	14
75%	668.500000	1.000000	3.000000	NaN	NaN		0.000000	NaN	3
4									•

train.groupby('Survived').mean()

	PassengerId	Pclass	Age	SibSp	Parch	Fare
Survived						
0	447.016393	2.531876	30.626179	0.553734	0.329690	22.117887
1	444.368421	1.950292	28.343690	0.473684	0.464912	48.395408

train.corr()

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare
Passengerld	1.000000	-0.005007	-0.035144	0.036847	-0.057527	-0.001652	0.012658
Survived	-0.005007	1.000000	-0.338481	-0.077221	-0.035322	0.081629	0.257307
Pclass	-0.035144	-0.338481	1.000000	-0.369226	0.083081	0.018443	-0.549500
Age	0.036847	-0.077221	-0.369226	1.000000	-0.308247	-0.189119	0.096067
SibSp	-0.057527	-0.035322	0.083081	-0.308247	1.000000	0.414838	0.159651
Parch	-0.001652	0.081629	0.018443	-0.189119	0.414838	1.000000	0.216225
Fare	0.012658	0.257307	-0.549500	0.096067	0.159651	0.216225	1.000000



alive=len(train[train['Survived']==1])

dead=len(train[train['Survived']==0])

train.groupby('Sex')[['Survived']].mean()

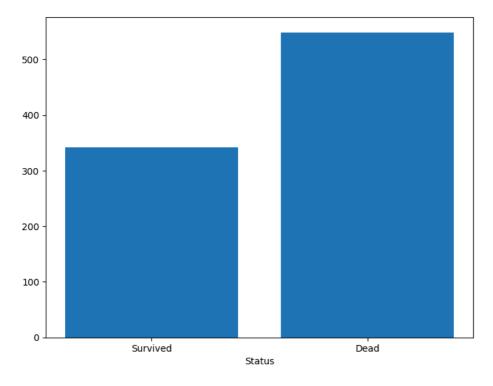
Survived

 Sex

 female
 0.742038

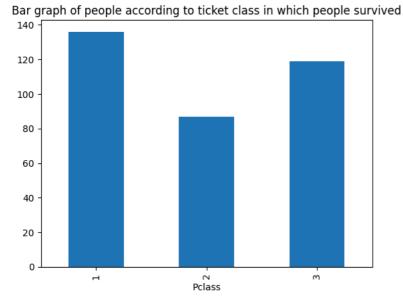
 male
 0.188908

fig=plt.figure()
ax=fig.add_axes([0,0,1,1])
Status=['Survived','Dead']
ind=[alive,dead]
ax.bar(Status,ind)
plt.xlabel('Status')
plt.show()



plt.figure(1)
train.loc[train['Survived'] == 1,'Pclass'].value_counts().sort_index().plot.bar()
plt.title("Bar graph of people according to ticket class in which people survived")

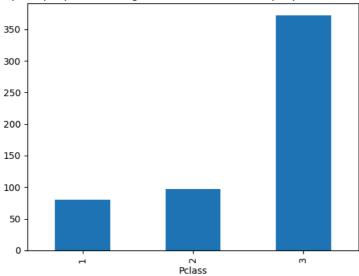
 ${\sf Text}({\tt 0.5},\,{\tt 1.0},\,{\tt 'Bar graph \ of people \ according \ to \ ticket \ class \ in \ which \ people \ survived')}$



plt.figure(2)
train.loc[train['Survived'] == 0,'Pclass'].value_counts().sort_index().plot.bar()
plt.title("Bar graph of people according to ticket class in which people could not survived")

 ${\sf Text}(\textbf{0.5, 1.0, 'Bar graph of people according to ticket class in which people could not survived')}$

Bar graph of people according to ticket class in which people could not survived



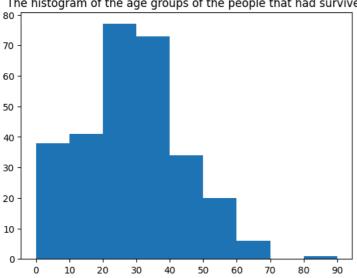
```
plt.figure(1)
age = train.loc[train.Survived == 1, 'Age']

plt.title("The histogram of the age groups of the people that had survived")
plt.hist(age,np.arange(0,100,10))
```

plt.xticks(np.arange(0,100,10))

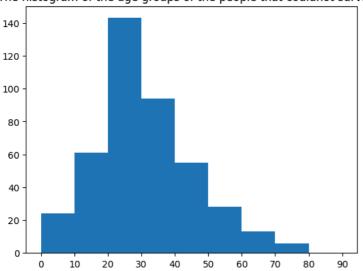
```
([<matplotlib.axis.XTick at 0x7e195df21ff0>,
  <matplotlib.axis.XTick at 0x7e195df22080>,
  <matplotlib.axis.XTick at 0x7e195df22800>,
  <matplotlib.axis.XTick at 0x7e195d9aaf20>,
  <matplotlib.axis.XTick at 0x7e195d9a8310>,
  <matplotlib.axis.XTick at 0x7e195d9ab370>,
  <matplotlib.axis.XTick at 0x7e195da30610>,
  <matplotlib.axis.XTick at 0x7e195d9a89d0>,
  <matplotlib.axis.XTick at 0x7e195da31030>,
  <matplotlib.axis.XTick at 0x7e195da31ae0>],
 [Text(0, 0, '0'),
Text(10, 0, '10'),
Text(20, 0, '20'),
  Text(30, 0, '30'),
 Text(40, 0, '40'),
Text(50, 0, '50'),
 Text(60, 0, '60'),
Text(70, 0, '70'),
Text(80, 0, '80'),
Text(90, 0, '90')])
```

The histogram of the age groups of the people that had survived



```
plt.figure(1)
age = train.loc[train.Survived == 0, 'Age']
plt.title("The histogram of the age groups of the people that couldnot survived")
plt.hist(age,np.arange(0,100,10))
plt.xticks(np.arange(0,100,10))
```

The histogram of the age groups of the people that couldnot survived



train[['SibSp','Survived']].groupby(['SibSp'],as_index=False).mean().sort_values(by='Survived',ascending=False)

	SibSp	Survived
1	1	0.535885
2	2	0.464286
0	0	0.345395
3	3	0.250000
4	4	0.166667
5	5	0.000000
6	8	0.000000

 $train[['Pclass', 'Survived']]. group by (['Pclass'], as_index=False). mean(). sort_values(by='Survived', ascending=False)) and the survived of the survived$

	Pclass	Survived
0	1	0.629630
1	2	0.472826
2	3	0.242363

 $\label{train[['Age','Survived']].groupby(['Age'],as_index=False).mean().sort_values(by='Age',ascending=False)} \\$

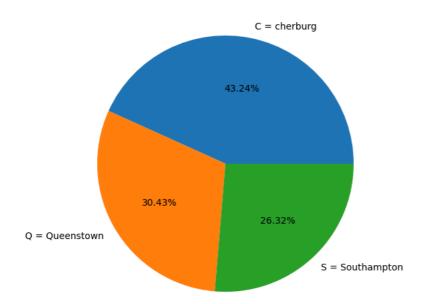
	Age	Survived
87	80.00	1.0
86	74.00	0.0
85	71.00	0.0
84	70.50	0.0
83	70.00	0.0
4	0.92	1.0
3	0.83	1.0
2	0.75	1.0
1	0.67	1.0
0	0.42	1.0

88 rows × 2 columns

 $\label{train} train[['Embarked', 'Survived']].groupby(['Embarked'], as_index=False).mean().sort_values(by='Survived', ascending=False)$

	Embarked	Survived
0	С	0.553571
1	Q	0.389610
2	S	0.336957

```
fig= plt.figure()
ax =fig.add_axes([0,0,1,1])
ax.axis('equal')
l = ['C = cherburg','Q = Queenstown','S = Southampton']
s =[0.553571,0.389610,0.336957]
ax.pie(s,labels= l,autopct='%1.2f%%')
plt.show()
```



test.describe(include='all')

```
PassengerId
                         Pclass
                                  Name
                                          Sex
                                                      Age
                                                                     Parch Ticket
                                                                               418 417.
count
         418.000000 418.000000
                                    418
                                          418
                                              332.000000
                                                                418.000000
unique
               NaN
                           NaN
                                    418
                                            2
                                                     NaN
                                                                      NaN
                                                                               363
                                  Kelly,
                                                                                PC
 top
               NaN
                           NaN
                                    Mr.
                                        male
                                                     NaN
                                                                      NaN
                                                                             17608
                                  James
 freq
                NaN
                           NaN
                                      1
                                          266
                                                     NaN
                                                                      NaN
                                                                                 5
        1100.500000
                       2.265550
                                   NaN
                                         NaN
                                                30.272590
                                                                  0.392344
                                                                               NaN
                                                                                     35.
mean
 std
         120.810458
                       0.841838
                                   NaN
                                         NaN
                                                14.181209
                                                                  0.981429
                                                                               NaN
                                                                                     55.
         892 000000
                       1 000000
                                   NaN
                                         NaN
                                                 0.170000
                                                                  0.000000
                                                                               NaN
                                                                                      0
 min
 25%
         996.250000
                       1.000000
                                   NaN
                                         NaN
                                                21.000000
                                                                  0.000000
                                                                               NaN
                                                                                       7.
 50%
                                                                                     14.
        1100.500000
                       3.000000
                                   NaN
                                         NaN
                                                27.000000
                                                                  0.000000
                                                                               NaN
 75%
        1204.750000
                       3.000000
                                   NaN
                                         NaN
                                                39.000000
                                                                  0.000000
                                                                               NaN
                                                                                     31.
```

```
train= train.drop(['Ticket'],axis=1)
test= test.drop(['Ticket'],axis=1)
train= train.drop(['Cabin'],axis=1)
test= test.drop(['Cabin'],axis=1)
train= train.drop(['Name'],axis=1)
test= test.drop(['Name'],axis=1)
#feutures selection
column_train=['Age','Pclass','SibSp','Parch','Fare','Sex','Embarked']
#training values
X=train[column_train]
#target value
Y=train['Survived']
X['Age'].isnull().sum()
X['Pclass'].isnull().sum()
X['SibSp'].isnull().sum()
X['Parch'].isnull().sum()
X['Fare'].isnull().sum()
X['Sex'].isnull().sum()
X['Embarked'].isnull().sum()
X['Age']=X['Age'].fillna(X['Age'].median())
X['Age'].isnull().sum()
     0
X['Embarked']=train['Embarked'].fillna(method='pad')
X['Embarked'].isnull().sum()
     0
d={'male':0,'female':1}
X['Sex']=X['Sex'].apply(lambda x:d[x])
X['Sex'].head()
```

```
2
     3
          1
     4
          0
     Name: Sex, dtype: int64
e={'C':0,'Q':1,'S':2}
X['Embarked']=X['Embarked'].apply(lambda x:e[x])
X['Embarked'].head()
          2
     0
          0
     1
     2
          2
     3
          2
     Name: Embarked, dtype: int64
from sklearn.model_selection import train_test_split
X_train,X_test,Y_train,Y_test=train_test_split(X,Y,test_size=0.3,random_state=7)
from \ sklearn.linear\_model \ import \ LogisticRegression
model= LogisticRegression()
model.fit(X_train,Y_train)
Y_pred=model.predict(X_test)
from sklearn.metrics import accuracy_score
print("Accuracy Score;",accuracy_score(Y_test,Y_pred))
     Accuracy Score; 0.753731343283582
from \ sklearn.metrics \ import \ accuracy\_score, confusion\_matrix
confusion_mat = confusion_matrix(Y_test,Y_pred)
print(confusion_mat)
     [[133 23]
      [ 43 69]]
from sklearn.svm import SVC
model1=SVC()
model1.fit(X_train,Y_train)
pred_y = model1.predict(X_test)
from sklearn metrics import accuracy score
```