In [1]:

import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings("ignore")

In [2]:

df=pd.read\_csv("https://raw.githubusercontent.com/dsrscientist/dataset1/mas
ter/titanic\_train.csv")

df

Out[2]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund,		22.0		0	A/5 21171	7.2500		S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	С
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/02. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000	NaN	S
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000	B42	S
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500	NaN	S
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000	C148	С

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
890	891	0		Dooley, Mr. Patrick	male	32.0	0	0	370376	7.7500	NaN	Q

891 rows × 12 columns

df.head()

In [3]:

										Out[3]:  Fare Cabin Embarke				
	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked		
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S		
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	С		
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/02. 3101282	7.9250	NaN	S		
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S		
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S		

In [4]:

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	PassengerId	891 non-null	int64
1	Survived	891 non-null	int64
2	Pclass	891 non-null	int64
3	Name	891 non-null	object
4	Sex	891 non-null	object
5	Age	714 non-null	float64
6	SibSp	891 non-null	int64
7	Parch	891 non-null	int64
8	Ticket	891 non-null	object
9	Fare	891 non-null	float64
10	Cabin	204 non-null	object
11	Embarked	889 non-null	object

dtypes: float64(2), int64(5), object(5)

memory usage: 83.7+ KB

df.isnull().sum()

In [5]:

Out[5]:

PassengerId 0 Survived 0 Pclass 0 Name 0 Sex 0 177 Age SibSp 0 Parch 0 Ticket 0 0 Fare 687 Cabin Embarked

dtype: int64

df.describe()

In [6]:

Out[6]:

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

In [7]:

# dropping Cabin as more than 50% Null values are available # replacing nul value of age with mean

In [8]:

df=df.drop(columns='Cabin',axis=1)
df['Age'].fillna(df['Age'].mean(), inplace=True)

df['Embarked'].fillna(df['Embarked'].mode()[0], inplace=True)

In [9]:

df.head()

Out[9]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	S
1	2	1	11	Cumings, Mrs. John Bradley (Florence	female	38.0	1	0	PC 17599	71.2833	С

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Embarked
				Briggs Th							
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/02. 3101282	7.9250	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	S
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	S

```
In [10]:
categorical=[]
for i in df.dtypes.index:
    if df.dtypes[i] == "object":
        categorical.append(i)
print("Categorical columns:",categorical)
print("\n")
numerical=[]
for i in df.dtypes.index:
    if df.dtypes[i]!="object":
        numerical.append(i)
print("Numerical columns:", numerical)
print("\n")
Categorical columns: ['Name', 'Sex', 'Ticket', 'Embarked']
Numerical columns: ['PassengerId', 'Survived', 'Pclass', 'Age', 'SibSp',
'Parch', 'Fare']
                                                                         In [11]:
df["Survived"].unique()
                                                                         Out[11]:
array([0, 1], dtype=int64)
                                                                         In [12]:
df["Ticket"].nunique()
                                                                        Out[12]:
681
                                                                         In [13]:
df["Embarked"].nunique()
                                                                        Out[13]:
3
                                                                         In [14]:
#Now we can replace Survived and Embarked to numerical column
```

```
In [15]:
df.replace({'Sex':{'male':0,'female':1}, 'Embarked':{'S':0,'C':1,'Q':2}},
inplace=True)
```

In [16]:

df.head()

plt.show()

Out[16]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Embarked
0	1	0	3	Braund, Mr. Owen Harris	0	22.0	1	0	A/5 21171	7.2500	0
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	1	38.0	1	0	PC 17599	71.2833	1
2	3	1	3	Heikkinen, Miss. Laina	1	26.0	0	0	STON/02. 3101282	7.9250	0
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	1	35.0	1	0	113803	53.1000	0
4	5	0	3	Allen, Mr. William Henry	0	35.0	0	0	373450	8.0500	0

In [17]:

dfnew=df.drop(['PassengerId','Name','Ticket'], axis=1)
dfnew.head()

Out[17]:

	Survived	<b>Pclass</b>	Sex	Age	SibSp	Parch	Fare	Embarked
0	0	3	0	22.0	1	0	7.2500	0
1	1	1	1	38.0	1	0	71.2833	1
2	1	3	1	26.0	0	0	7.9250	0
3	1	1	1	35.0	1	0	53.1000	0
4	0	3	0	35.0	0	0	8.0500	0

In [18]:

```
plt.figure(figsize=(15,10))
sns.heatmap(dfnew.corr())
sns.heatmap(dfnew.corr(), annot = True, vmin=-1, vmax=1, center= 0, cmap=
'Blues', linewidths=3, linecolor='black')
```

Survived	1	-0.34	0.54	-0.07	-0.035	0.082	0.26	0
Pclass	-0.34	1	-0.13	-0.33	0.083	0.018	-0.55	0.0
Sex	0.54	-0.13	1	-0.084	0.11	0.25	0.18	0.
Age	-0.07	-0.33	-0.084	1	-0.23	-0.18	0.092	o.c
SibSp	-0.035	0.083	0.11	-0.23	1	0.41	0.16	-0
Parch	0.082	0.018	0.25	-0.18	0.41	1	0.22	-0.
Fare	0.26	-0.55	0.18	0.092	0.16	0.22	1	0.0
Embarked	0.11	0.046	0.12	0.0075	-0.06	-0.079	0.062	
'	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Emb

# Spliting of data into training(80%) and testing(20%) sets

from sklearn.model\_selection import train\_test\_split

In [20]:

In [19]:

<sup>#</sup>taking high correlation value

<sup>#</sup> X = features, y = target variable

```
X = dfnew[['Pclass','Sex','Age','SibSp','Parch','Fare','Embarked']]
y = dfnew['Survived']
X train, X test, y train, y test = train test split(X, y, test size = 0.2,
random state = 0)
                                                                          In [21]:
from sklearn.preprocessing import StandardScaler
scale=StandardScaler()
dfnew=scale.fit transform(dfnew)
                                                                          In [22]:
from sklearn.linear model import LogisticRegression
                                                                          In [23]:
lr = LogisticRegression()
lr.fit(X_train,y_train)
                                                                         Out[23]:
LogisticRegression()
                                                                          In [24]:
y pred=lr.predict(X test)
actual vs pred=pd.DataFrame({'Actual': y test, 'Predicted': y pred})
actual_vs_pred
                                                                         Out[24]:
    Actual Predicted
495 0
           0
648 0
278 0
           0
 31 1
           1
255 1
           1
 ...
780 1
           1
837 0
           0
215 1
           1
833 0
           0
372 0
          0
179 rows × 2 columns
                                                                           In []:
                                                                           In []:
                                                                          In [36]:
from sklearn.metrics import accuracy score
                                                                          In [37]:
X test prediction=lr.predict(X test)
test data accuracy=accuracy score(y test, X test prediction)
print('Accuracy score of test data : ', test_data_accuracy)
Accuracy score of test data: 0.8044692737430168
                                                                           In []:
```

In []:

In []: