LOGISTIC REGRESSION

PROBLEM STATEMENT: TO PREDICT AND ANALYZE WHICH GENDER HAS A HIGH CHANCE OF SURVIVAL AT THE TIME OF DISASTER

In [1]:

```
import numpy as np
import pandas as pd
from sklearn import preprocessing
import matplotlib.pyplot as plt
import seaborn as sns
sns.set(style="white")
sns.set(style="whitegrid",color_codes=True)
import warnings
warnings.simplefilter(action='ignore')
```

In [2]:

 $\label{train_df} train_df=pd.read_csv(r"C:\Users\91720\Downloads\train.gender_submission.csv") train_df$

Out[2]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fa
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.25
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.28
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.92
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.10
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.05
	•••									
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.00
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.00
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.45
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.00
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.75

891 rows × 12 columns

In [3]:

test_df=pd.read_csv(r"C:\Users\91720\Downloads\test.gender_submission.csv")
test_df

Out[3]:

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cat
0	892	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292	N
1	893	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000	N
2	894	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	N
3	895	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	N
4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	N
413	1305	3	Spector, Mr. Woolf	male	NaN	0	0	A.5. 3236	8.0500	N
414	1306	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	108.9000	C1
415	1307	3	Saether, Mr. Simon Sivertsen	male	38.5	0	0	SOTON/O.Q. 3101262	7.2500	N
416	1308	3	Ware, Mr. Frederick	male	NaN	0	0	359309	8.0500	N
417	1309	3	Peter, Master. Michael J	male	NaN	1	1	2668	22.3583	N

418 rows × 11 columns

In [4]:

train_df.head()

Out[4]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500
4										•

In [5]:

train_df.shape

Out[5]:

(891, 12)

In [6]:

test_df.head()

Out[6]:

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Em
0	892	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292	NaN	
1	893	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000	NaN	
2	894	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	NaN	
3	895	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	NaN	
4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	NaN	
4		_	_	_	-	_	_	_	_		

In [7]:

test_df.shape

Out[7]:

(418, 11)

In [8]:

train_df.describe()

Out[8]:

	Passengerld	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 [9]:

train_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
d+vn	as: flaat64(2) int64(5) ohi	oct(5)

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

memory usage: 83.7+ KB

In [10]:

test_df.describe()

Out[10]:

	Passengerld	Pclass	Age	SibSp	Parch	Fare
count	418.000000	418.000000	332.000000	418.000000	418.000000	417.000000
mean	1100.500000	2.265550	30.272590	0.447368	0.392344	35.627188
std	120.810458	0.841838	14.181209	0.896760	0.981429	55.907576
min	892.000000	1.000000	0.170000	0.000000	0.000000	0.000000
25%	996.250000	1.000000	21.000000	0.000000	0.000000	7.895800
50%	1100.500000	3.000000	27.000000	0.000000	0.000000	14.454200
75%	1204.750000	3.000000	39.000000	1.000000	0.000000	31.500000
max	1309.000000	3.000000	76.000000	8.000000	9.000000	512.329200

```
In [11]:
```

```
test_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 418 entries, 0 to 417
Data columns (total 11 columns):
## Column
```

#	Column	Non-Null Count	Dtype
0	PassengerId	418 non-null	int64
1	Pclass	418 non-null	int64
2	Name	418 non-null	object
3	Sex	418 non-null	object
4	Age	332 non-null	float64
5	SibSp	418 non-null	int64
6	Parch	418 non-null	int64
7	Ticket	418 non-null	object
8	Fare	417 non-null	float64
9	Cabin	91 non-null	object
10	Embarked	418 non-null	object
44	C1+C4/2	\ :	+ / - \

dtypes: float64(2), int64(4), object(5)

memory usage: 36.1+ KB

TO FIND MISSION VALUES

In [12]:

```
test_df.isnull().sum()
```

Out[12]:

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

dtype: int64

In [13]:

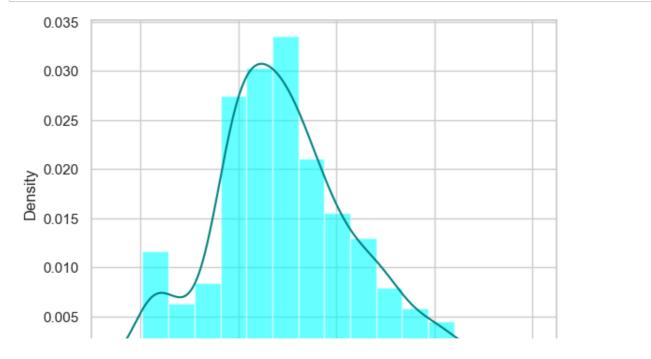
```
train_df.isnull().sum()
```

Out[13]:

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

In [14]:

```
ax=train_df['Age'].hist(bins=15,density=True,stacked=True,color='cyan',alpha=0.6)
train_df['Age'].plot(kind='density',color='teal')
ax.set(xlabel='Age')
plt.xlim(-10,85)
plt.show()
```



In [15]:

```
print(train_df['Age'].mean(skipna=True))
print(train_df['Age'].median(skipna=True))
```

29.69911764705882

28.0

In [16]:

```
print((train_df['Cabin'].isnull().sum()/train_df.shape[0])*100)
```

77.10437710437711

In [17]:

```
print((train_df['Embarked'].isnull().sum()/train_df.shape[0])*100)
```

0.22446689113355783

In [18]:

```
print('Boarded passengers grouped by port of embarkantion(c=Cherbourg,q=Queenstown,s=Sour
print(train_df['Embarked'].value_counts())
sns.countplot(x='Embarked',data=train_df,palette='Set2')
plt.show()
```

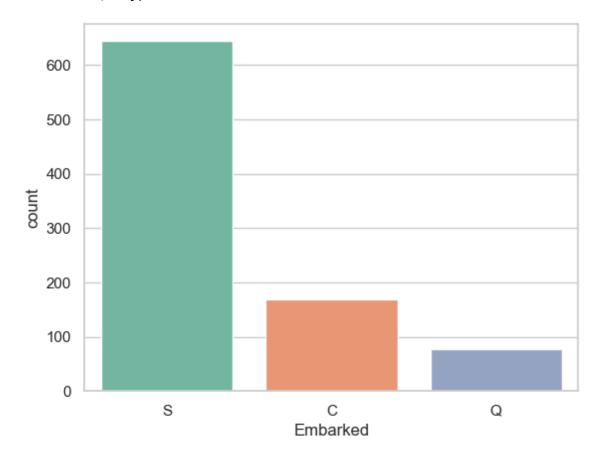
Boarded passengers grouped by port of embarkantion(c=Cherbourg,q=Queenstow n,s=Southampton):

Embarked S 644 C 168

77

Q

Name: count, dtype: int64



In [19]:

```
print(train_df['Embarked'].value_counts().idxmax())
```

In [20]:

```
train_data=train_df.copy()
train_data['Age'].fillna(train_df["Age"].median(skipna=True),inplace=True)
train_data["Embarked"].fillna(train_df['Embarked'].value_counts().idxmax(),inplace=True)
train_data.drop('Cabin',axis=1,inplace=True)
```

In [21]:

train_data.isnull().sum()

Out[21]:

0
0
0
0
0
0
0
0
0
0
0

In [22]:

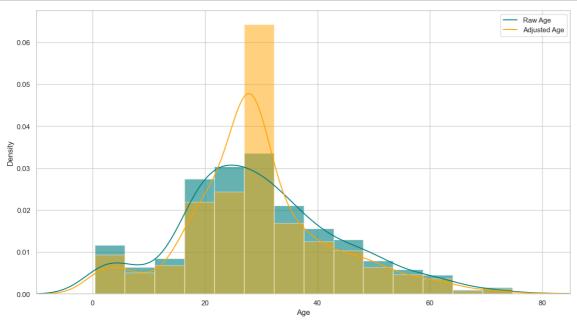
train_data.head()

Out[22]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500
4				_						

In [23]:

```
plt.figure(figsize=(15,8))
ax=train_df["Age"].hist(bins=15,density=True,stacked=True,color='teal',alpha=0.6)
train_df["Age"].plot(kind='density',color='teal')
ax=train_data["Age"].hist(bins=15,density=True,stacked=True,color='orange',alpha=0.5)
train_data["Age"].plot(kind='density',color='orange')
ax.legend(['Raw Age','Adjusted Age'])
ax.set(xlabel='Age')
plt.xlim(-10,85)
plt.show()
```



In [27]:

```
train_data['TravelAlone']=np.where((train_data["SibSp"]+train_data["Parch"])>0,0,1)
train_data.drop('SibSp',axis=1,inplace=True)
train_data.drop('Parch',axis=1,inplace=True)
```

In [28]:

```
training=pd.get_dummies(train_data, columns=["Pclass","Embarked","Sex"])
training.drop('Sex_female', axis=1, inplace=True)
training.drop('PassengerId', axis=1, inplace=True)
training.drop('Name', axis=1, inplace=True)
training.drop('Ticket', axis=1, inplace=True)
final_train = training
final_train.head()
```

Out[28]:

	Survived	Age	Fare	TravelAlone	Pclass_1	Pclass_2	Pclass_3	Embarked_C	Embark
0	0	22.0	7.2500	0	False	False	True	False	
1	1	38.0	71.2833	0	True	False	False	True	
2	1	26.0	7.9250	1	False	False	True	False	
3	1	35.0	53.1000	0	True	False	False	False	
4	0	35.0	8.0500	1	False	False	True	False	
4		_							

In [29]:

```
test_df.isnull().sum()
```

Out[29]:

0
0
0
0
86
0
0
0
1
327
0

In [30]:

```
test_data = test_df.copy()
test_data["Age"].fillna(train_df["Age"].median(skipna=True), inplace=True)
test_data["Fare"].fillna(train_df["Fare"].median(skipna=True), inplace=True)
test_data.drop('Cabin', axis=1, inplace=True)
test_data['TravelAlone']=np.where((test_data["SibSp"]+test_data["Parch"])>0, 0, 1)
test_data.drop('SibSp', axis=1, inplace=True)
test_data.drop('Parch', axis=1, inplace=True)
testing = pd.get_dummies(test_data, columns=["Pclass","Embarked","Sex"])
testing.drop('Sex_female', axis=1, inplace=True)
testing.drop('PassengerId', axis=1, inplace=True)
testing.drop('Name', axis=1, inplace=True)
testing.drop('Ticket', axis=1, inplace=True)
final_test = testing
final_test.head()
```

Out[30]:

	Age	Fare	TravelAlone	Pclass_1	Pclass_2	Pclass_3	Embarked_C	Embarked_Q	Em
0	34.5	7.8292	1	False	False	True	False	True	
1	47.0	7.0000	0	False	False	True	False	False	
2	62.0	9.6875	1	False	True	False	False	True	
3	27.0	8.6625	1	False	False	True	False	False	
4	22.0	12.2875	0	False	False	True	False	False	
4									

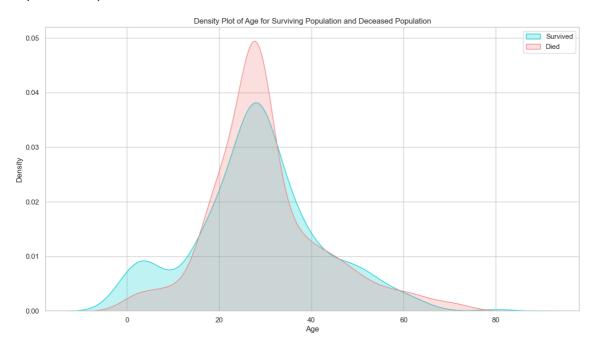
EXPLORATORY DATA ANALYSIS

In [31]:

```
plt.figure(figsize=(15,8))
ax = sns.kdeplot(final_train["Age"][final_train.Survived == 1], color="darkturquoise",sh
sns.kdeplot(final_train["Age"][final_train.Survived == 0], color="lightcoral", shade=Tru
plt.legend(['Survived', 'Died'])
plt.title('Density Plot of Age for Surviving Population and Deceased Population')
```

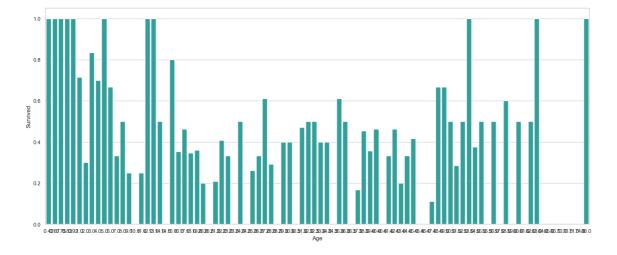
Out[31]:

Text(0.5, 1.0, 'Density Plot of Age for Surviving Population and Deceased Population')



In [35]:

```
plt.figure(figsize=(20,8))
avg_survival_byage = final_train[["Age", "Survived"]].groupby(['Age'], as_index=False).m
g = sns.barplot(x='Age', y='Survived', data=avg_survival_byage, color="LightSeaGreen")
plt.show()
```



```
In [36]:
```

Name: IsMinor, Length: 418, dtype: int32

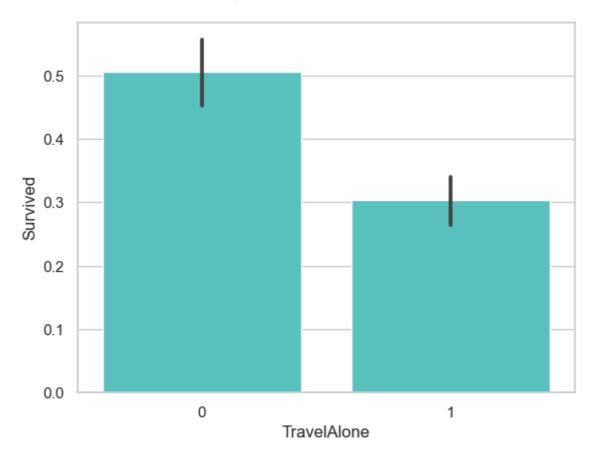
```
final_train['IsMinor']=np.where(final_train['Age']<=16, 1, 0)</pre>
print(final_train['IsMinor'])
0
       0
1
       0
2
       0
3
       0
4
       0
886
       0
887
       0
       0
888
889
       0
890
       0
Name: IsMinor, Length: 891, dtype: int32
In [37]:
final_test['IsMinor']=np.where(final_test['Age']<=16, 1, 0)</pre>
print(final_test['IsMinor'])
0
       0
1
       0
2
       0
3
       0
4
       0
413
       0
414
       0
415
       0
416
       0
```

In [38]:

sns.barplot(x='TravelAlone', y='Survived', data=final_train, color="mediumturquoise")

Out[38]:

<Axes: xlabel='TravelAlone', ylabel='Survived'>



In [39]:

```
import seaborn as sns
import matplotlib.pyplot as plt
sns.barplot(x='Sex',y='Survived',data=train_df,color='aquamarine')
plt.show()
```

