LOGISTIC REGRESSION PROCESS:

problem statement:- ¶

To predict and Analysis which gender has a high chance of survival at the time of disaster

*import Datasets,python packages and libraries

In [1]:

```
import numpy as np
import pandas as pd
from sklearn import preprocessing
import matplotlib.pyplot as plt
#plt.rc("font", size=14)
import seaborn as sns
sns.set(style="white")#white background style for seaborn plots
sns.set(style="whitegrid",color_codes=True)
import warnings
warnings.simplefilter(action='ignore')
```

In [2]:

 $\label{train_df} train_df=pd.read_csv(r"C:\Users\shaha\OneDrive\Desktop\Excel\train.gender_submission.csv\\ train_df$

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	8.050
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 r	ows × 12 colu	ımns								
4										•

In [3]:

 $test_df=pd.read_csv(r"C:\Users\shaha\OneDrive\Desktop\Excel\test_gender_submission.csv")\\ test_df$

Out[3]:

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cal
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

localhost:8888/notebooks/Logistic Regression (gender of train %26 test model).ipynb

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]:

 0 1 2 3 4 886 887 888 889 890	nd metho	NDFrame.0 1 2 3 4 5 887 888 889 890 891	describ	e of 3 \ 1 3 2 1 3 1 3 3	Passer	ngerId	Sur	vived	Pclass	
р						N	ame	Sex	Age	SibS
0				Braund	, Mr. Ow	en Har	ris	male	22.0	
1 \ 1	Cumings	, Mrs. Joh	n Bradl	ey (Flor	ence Bri	.ggs Th		female	38.0	
1 2				Hoikk	inen, Mi	cc la	ina	female	26.0	
0					-					
3 1	Fu	trelle, Mrs	s. Jacq	ues Heat	n (Lily	May Pe	el)	female	35.0	
4				Allen, I	Mr. Will	iam He	nry	male	35.0	
0										
 886				Mont	√ila, Re	ev. Juo	zas	male	27.0	
0 887			Gra	ham, Mis	-					
0 888		Johnston		-						
1		Joinis Con	, MISS.	Catherin	іе петег	i Carr	те	тешате	. NaN	
889 0				Behr	, Mr. Ka	ırl How	ell	male	26.0	
890 0				Do	oley, Mr	. Patr	ick	male	32.0	
Ū		_	-• • •	_						
0	Parch 0		Γicket 21171	7.2500	Cabin E NaN		a S			
1	0		17599	71.2833	C85		C			
2	0	STON/02. 33		7.9250	NaN		S			
3	0		113803	53.1000	C123		S			
4	0	-	373450	8.0500	NaN		S			
 886	0	;	 211536	13.0000	 NaN	• •	· S			
887	0		112053	30.0000	B42		S			
888	2	W./C	. 6607	23.4500	NaN		S			
889	0		111369	30.0000	C148		C			
890	0	=	370376	7.7500	NaN		Q			
[891	rows x	12 columns]>							

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
	C1 1 C 4 / 2	\ '''	

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

memory usage: 83.7+ KB

In [10]:

test_df.describe

Out[10]:

<bou< th=""><th></th><th>od NDFr</th><th>ame.des</th><th>cribe of</th><th>PassengerId</th><th>Pclass</th><th></th></bou<>		od NDFr	ame.des	cribe of	PassengerId	Pclass	
0 1 2 3 4		892 893 894 895 896	3 3 2 3 3	Hirvoner	Wilkes, Mrs. 3 Myles, n, Mrs. Alexander (James (Elle Mr. Thomas Wirz, Mr	Francis Albert
••		•••	•••	HIII VOHEI	i, rii 3. Alexander (Tierga L Li	•••
413 414 415 416 417		1305 1306 1307 1308 1309	3 1 3 3 3		Saether, N	Spector, Mana, Dona. Mr. Simon Solare, Mr. Foliater, Mr. Foliater, Mr. Foliater, Master, Mast	Fermina ivertsen rederick
	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin Embark
ed 0 Q	male	34.5	0	0	330911	7.8292	NaN
1	female	47.0	1	0	363272	7.0000	NaN
S 2	male	62.0	0	0	240276	9.6875	NaN
Q 3 S	male	27.0	0	0	315154	8.6625	NaN
4 S	female	22.0	1	1	3101298	12.2875	NaN
••	• • •			• • •			•••
413 S	male	NaN	0	0	A.5. 3236	8.0500	NaN
414 C	female	39.0	0	0	PC 17758	108.9000	C105
415 S	male	38.5	0	0 9	SOTON/O.Q. 3101262	7.2500	NaN
416	male	NaN	0	0	359309	8.0500	NaN
S 417 C	male	NaN	1	1	2668	22.3583	NaN

[418 rows x 11 columns]>

In [11]:

```
test_df.info()
```

<class 'pandas.core.frame.DataFrame'> RangeIndex: 418 entries, 0 to 417 Data columns (total 11 columns): Column Non-Null Count Dtype -----0 PassengerId 418 non-null int64 1 Pclass 418 non-null int64 2 Name object 418 non-null object 3 Sex 418 non-null 4 Age 332 non-null float64 5 418 non-null int64 SibSp 6 Parch 418 non-null int64 7 418 non-null Ticket object 8 Fare 417 non-null float64 9 91 non-null object Cabin

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

418 non-null

memory usage: 36.0+ KB

10 Embarked

TO FIND MISSING VALUES

In [12]:

train_df.isnull().sum()

object

Out[12]:

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

In [13]:

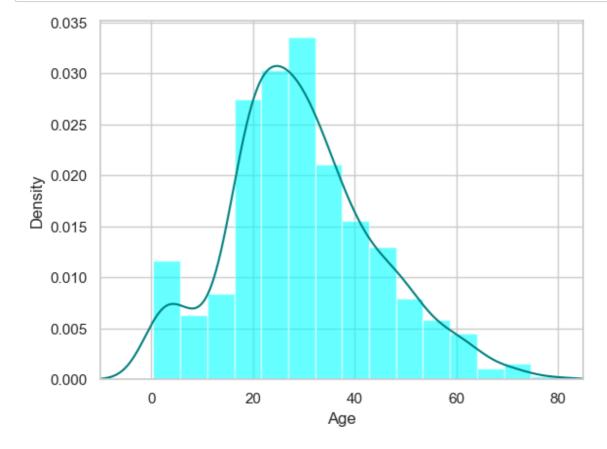
```
test_df.isnull().sum()
```

Out[13]:

PassengerId 0 **Pclass** 0 Name 0 Sex 0 Age 86 SibSp 0 Parch 0 Ticket 0 Fare 1 Cabin 327 Embarked 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 part of embarkation(c=Cherbourg,Q=Queenstown,S=Sout
print(train_df['Embarked'].value_counts())
sns.countplot(x='Embarked',data=train_df)
plt.show()
```

Boarded passengers grouped by part of embarkation(c=Cherbourg,Q=Queenstow n,S=Southampton:)

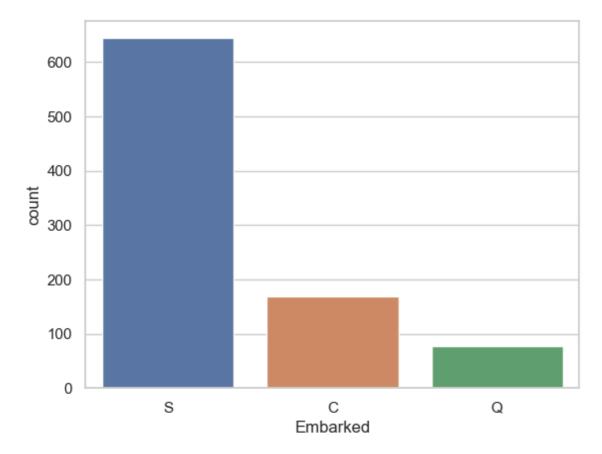
Embarked

S 644

C 168

Q 77

Name: count, dtype: int64



In [19]:

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

S

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]:

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

In [22]:

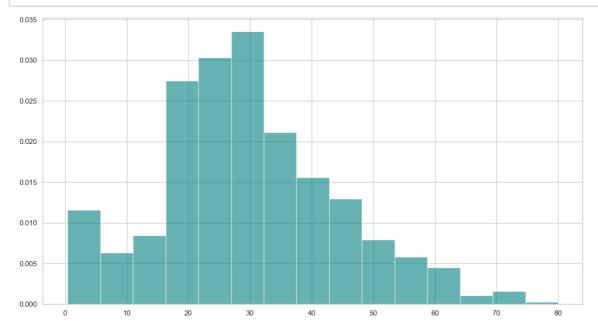
train_data.head()

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										•

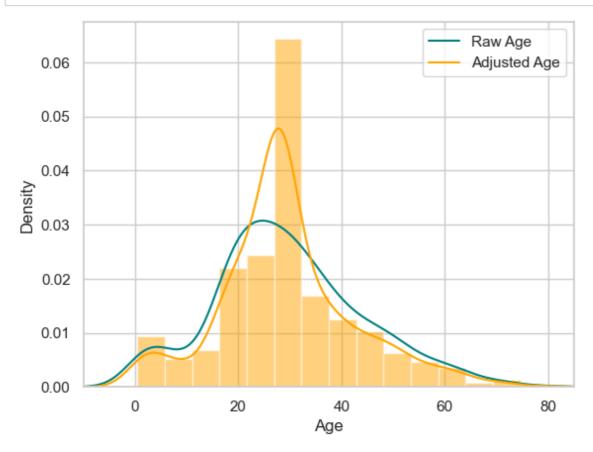
In [23]:

```
plt.figure(figsize=(15,8))
ax=train_df["Age"].hist(bins=15,density=True,stacked=True,color='teal',alpha=0.6)
```



In [24]:

```
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 [25]:

```
#create categorical variable for travelling alone
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 [26]:

```
#create categorical variables and drop some variables
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[26]:

	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 [27]:

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

Out[27]:

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 [28]:

```
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('Ticket',axis=1,inplace=True)
final_test=testing
final_test.head()
```

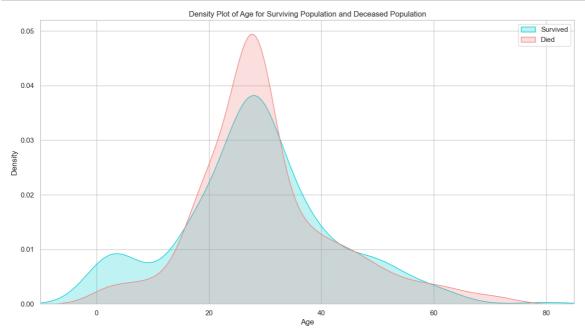
Out[28]:

	Name	Age	Fare	TravelAlone	Pclass_1	Pclass_2	Pclass_3	Embarked_C	Embarl
0	Kelly, Mr. James	34.5	7.8292	1	False	False	True	False	
1	Wilkes, Mrs. James (Ellen Needs)	47.0	7.0000	0	False	False	True	False	
2	Myles, Mr. Thomas Francis	62.0	9.6875	1	False	True	False	False	
3	Wirz, Mr. Albert	27.0	8.6625	1	False	False	True	False	
4	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	22.0	12.2875	0	False	False	True	False	
4									•

EXPLORATORY DATA ANALYSIS

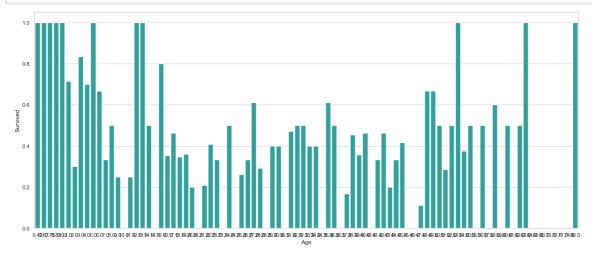
In [32]:

```
plt.figure(figsize=(15,8))
ax=sns.kdeplot(final_train["Age"][final_train.Survived == 1], color="darkturquoise", sha
sns.kdeplot(final_train["Age"][final_train.Survived == 0],color="lightcoral", shade=True
plt.legend(['Survived','Died'])
plt.title('Density Plot of Age for Surviving Population and Deceased Population')
ax.set(xlabel='Age')
plt.xlim(-10,85)
plt.show()
```



In [35]:

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

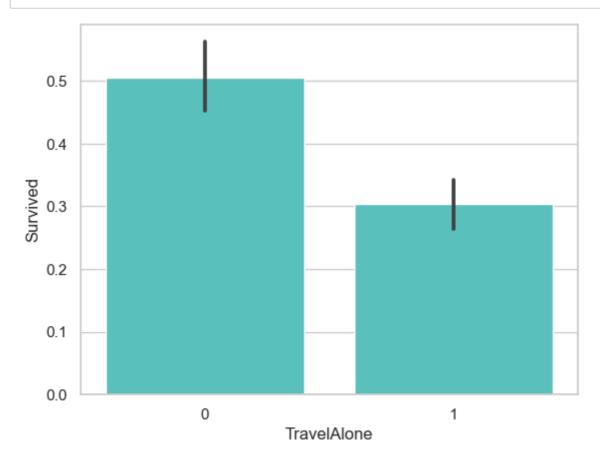


In [65]:

```
final_train['IsMinor']=np.where(final_train['Age']<=16,1,0)</pre>
print(final_train['IsMinor'])
0
       0
       0
1
2
       0
3
       0
       0
4
886
       0
887
       0
888
       0
       0
889
890
       0
Name: IsMinor, Length: 891, dtype: int32
In [66]:
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
       0
415
416
       0
417
Name: IsMinor, Length: 418, dtype: int32
```

In [67]:

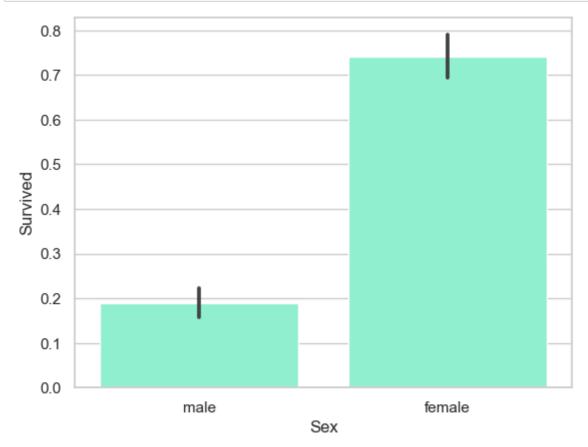
sns.barplot(x='TravelAlone',y='Survived',data=final_train,color="mediumturquoise")
plt.show()



In [68]:

```
import seaborn as sns
import matplotlib.pyplot as plt

#Assuming 'train_df' is your DataFrame containing the data
sns.barplot(x='Sex',y='Survived',data=train_df,color='aquamarine')
plt.show()
```



In []: