In [1]:

```
import pandas as pd
import numpy as np
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]:

- train_df=pd.read_csv(r"C:\Users\91955\Downloads\train.gender_submission.csv")
- train_df

Out[2]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000	NaN
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000	B42
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500	NaN
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000	C148
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.7500	NaN

891 rows × 12 columns

In [3]:

1 test_df=pd.read_csv(r"C:\Users\91955\Downloads\test.gender_submission.csv")

2 test_df

Out[3]:

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Emba
0	892	3	Kelly, Mr. James	ma l e	34.5	0	0	330911	7.8292	NaN	
1	893	3	Wilkes, Mrs. James (Ellen Needs)	fema l e	47.0	1	0	363272	7.0000	NaN	
2	894	2	Myles, Mr. Thomas Francis	ma l e	62.0	0	0	240276	9.6875	NaN	
3	895	3	Wirz, Mr. Albert	ma l e	27.0	0	0	315154	8.6625	NaN	
4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	fema l e	22.0	1	1	3101298	12.2875	NaN	
413	1305	3	Spector, Mr. Woo l f	ma l e	NaN	0	0	A.5. 3236	8.0500	NaN	
414	1306	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	108.9000	C105	
415	1307	3	Saether, Mr. Simon Sivertsen	male	38.5	0	0	SOTON/O.Q. 3101262	7.2500	NaN	
416	1308	3	Ware, Mr. Frederick	ma l e	NaN	0	0	359309	8.0500	NaN	
417	1309	3	Peter, Master. Michael J	ma l e	NaN	1	1	2668	22.3583	NaN	

418 rows × 11 columns

In [4]:

1 train_df.shape

Out[4]:

(891, 12)

In [5]:

1 train_df.head()

Out[5]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin I
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	fema l e	38.0	1	0	PC 17599	71.2833	C85
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN

In [6]:

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
dtyp	es: float64(2), int64(5), obj	ect(5)

memory usage: 83.7+ KB

localhost:8888/notebooks/Downloads/python2023/Logistic Regression3.ipynb

In [7]:

1 train_df.describe()

Out[7]:

	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 [8]:

1 test_df.shape

Out[8]:

(418, 11)

In [9]:

1 test_df.head()

Out[9]:

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

```
In [10]:
```

```
1 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	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
dtyp	es: float64(2), int64(4), obj	ect(5)

memory usage: 36.0+ KB

In [11]:

```
1 test_df.describe()
```

Out[11]:

	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 [12]:

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

Out[12]:

0
0
0
0
0
177
0
0
0
0
687
2

In [13]:

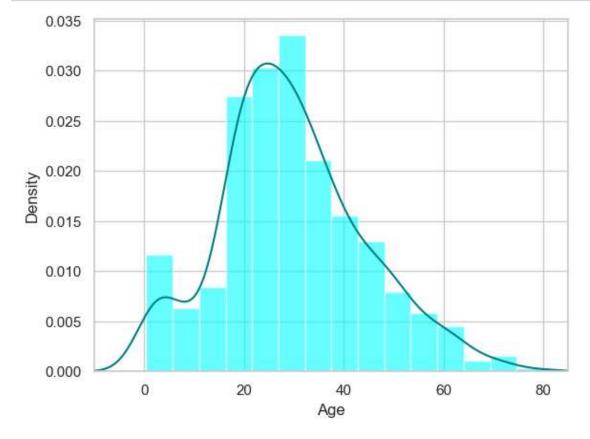
```
1 test_df.isnull().sum()
```

Out[13]:

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

In [14]:

```
1 ax=train_df["Age"].hist(bins=15,density=True,stacked=True,color='cyan',alpha=0.6)
2 train_df["Age"].plot(kind='density',color='teal')
3 ax.set(xlabel='Age')
4 plt.xlim(-10,85)
5 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)
```

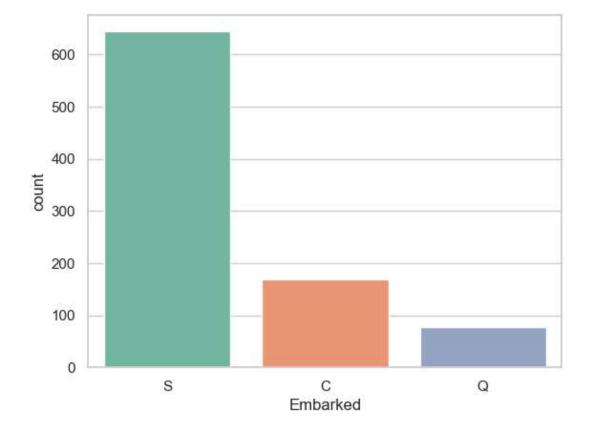
0.22446689113355783

In [18]:

```
print('Boarded passengers grouped by port of embarkation (C=Cherbourg,Q=Queentown,S=Southam;
print(train_df['Embarked'].value_counts())
sns.countplot(x='Embarked',data=train_df,palette='Set2')
plt.show()
```

Boarded passengers grouped by port of embarkation (C=Cherbourg,Q=Queentown,S=Sout hampton):
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]:

1 train_data.isnull().sum()

Out[21]:

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

In [22]:

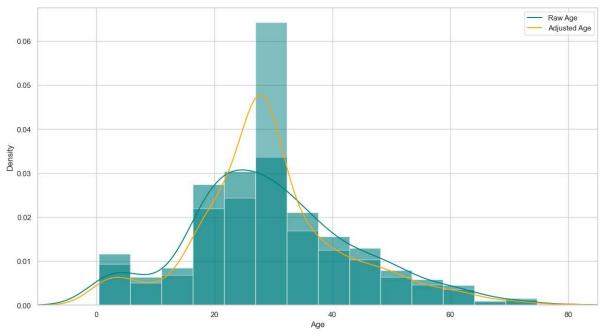
1 train_data.head()

Out[22]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Embarke
0	1	0	3	Braund, Mr. Owen Harris	ma l e	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	ma l e	35.0	0	0	373450	8.0500	

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='teal',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 [24]:

```
#categorical variable for traveling 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 [25]:

```
#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[25]:

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

In [26]:

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

Out[26]:

0
0
0
0
86
0
0
0
1
327
0

In [27]:

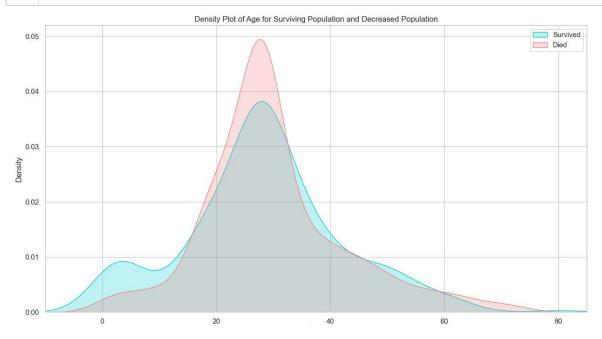
```
test data=test df.copy()
   test_data["Age"].fillna(train_df["Age"].median(skipna=True),inplace=True)
   test_data["Fare"].fillna(train_df["Fare"].median(skipna=True),inplace=True)
   test_data.drop('Cabin',axis=1,inplace=True)
 4
 5
 6
   test data['TravelAlone']=np.where((test data["SibSp"]+test data["Parch"])>0,0,1)
 7
   test data.drop('SibSp',axis=1,inplace=True)
   test_data.drop('Parch',axis=1,inplace=True)
 8
 9
10
   testing=pd.get_dummies(test_data,columns=["Pclass","Embarked","Sex"])
   testing.drop('Sex_female',axis=1,inplace=True)
11
12
   testing.drop('PassengerId',axis=1,inplace=True)
   testing.drop('Name',axis=1,inplace=True)
13
   testing.drop('Ticket',axis=1,inplace=True)
14
15
   final_test=testing
   final test.head()
```

Out[27]:

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

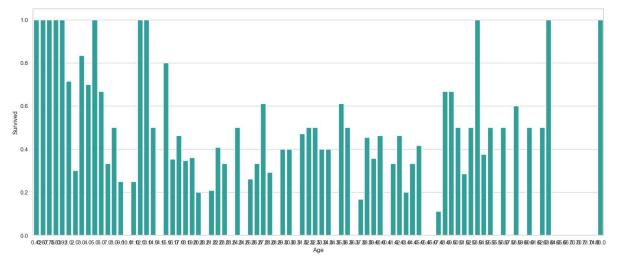
In [39]:

```
plt.figure(figsize=(15,8))
ax = sns.kdeplot(final_train["Age"][final_train.Survived == 1], color="darkturquoise",shade=
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 Decreased Population')
ax.set(xlabel='Age')
plt.xlim(-10,85)
plt.show()
```



In [37]:

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



In [31]:

```
final_train['IsMinor']=np.where(final_train['Age']<=16, 1, 0)
print(final_train['IsMinor'])</pre>
```

```
0
        0
        0
1
2
        0
3
        0
4
        0
886
        0
        0
887
888
        0
        0
889
890
Name: IsMinor, Length: 891, dtype: int32
```

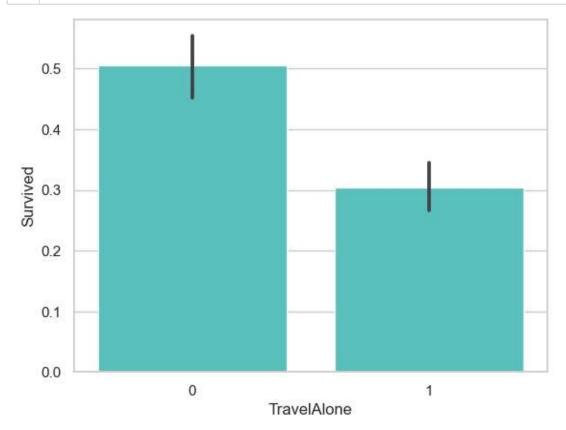
In [32]:

```
final_test['IsMinor']=np.where(final_test['Age']<=16, 1, 0)
print(final_test['IsMinor'])</pre>
```

```
0
        0
        0
1
2
        0
3
        0
4
        0
413
        0
414
        0
        0
415
416
        0
417
        0
Name: IsMinor, Length: 418, dtype: int32
```

In [34]:

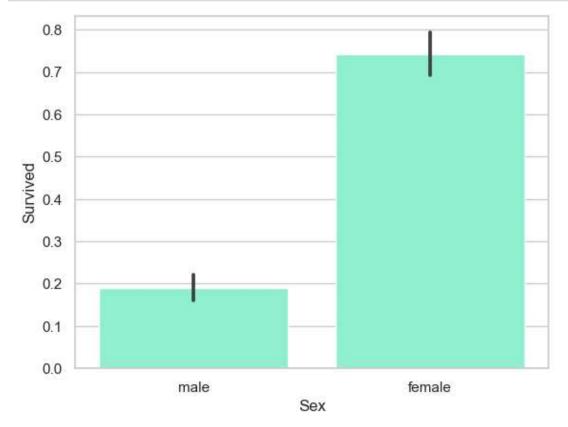
```
sns.barplot(x='TravelAlone', y='Survived', data=final_train, color="mediumturquoise")
plt.show()
```



In [35]:

```
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 []:

1