In [55]:

#To predict and analyse which gender has high chance of servival at the time of disaster
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='white')
sns.set(style='whitegrid',color_codes=True)
import warnings
warnings.simplefilter(action='ignore')

In [2]:

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

Out[2]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	С
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	
886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000	
887	888	1	1	Graham, Miss. Margaret Edith	female	19.0	0	0	112053	30.0000	
888	889	0	3	Johnston, Miss. Catherine Helen "Carrie"	female	NaN	1	2	W./C. 6607	23.4500	
889	890	1	1	Behr, Mr. Karl Howell	male	26.0	0	0	111369	30.0000	(
890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.7500	

891 rows × 12 columns

In [3]:

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

Out[3]:

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	E
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	
								•••			
413	1305	3	Spector, Mr. Woolf	male	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	male	NaN	0	0	359309	8.0500	NaN	
417	1309	3	Peter, Master. Michael J	male	NaN	1	1	2668	22.3583	NaN	

418 rows × 11 columns

In [4]:

train_df.head()

Out[4]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cab
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	Nε
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	Cł
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	Ne
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C1:
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	Ne
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	Embarke
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]:

train_df.shape

Out[7]:

(891, 12)

In [8]:

```
train_df.describe
```

Out[8]:

0 1 2 3 4	nd metho	od NDFramo 1 2 3 4 5	e.describe 0 1 1 1 0	3 1 3 1 3	Passeng	erId Su	ırvived	Pclass	\	
 886		887	0	2						
887		888	1	1						
888		889	0	3						
889		890	1	1						
890		891	0	3						
0 1 2 3 4	_	s, Mrs. Jo utrelle, M		ey (Flore Heikki ues Heath	nen, Mis	gs Th s. Laina ay Peel)	female female female female	22.0 38.0 26.0 35.0	SibSp 1 1 0 1	\
886 887 888 889 890		Johnsto	Gra on, Miss.	ham, Miss Catherin Behr,		et Edith "Carrie" l Howell	female female male	19.0 NaN 26.0	 0 0 1 0	
	Parch		Ticket	Fare	Cabin Em	barked				
0	0	A	/5 21171	7.2500	NaN	S				
1	0	I	PC 17599	71.2833	C85	С				
2	0	STON/02.		7.9250	NaN	S				
3	0		113803	53.1000	C123	S				
4	0		373450	8.0500	NaN	S				
••	• • • •		211526	12 0000	· · ·	• • • •				
886 887	0 0		211536 112053	13.0000 30.0000	NaN B42	S S				
888	2	لما	/C. 6607	23.4500	NaN	5 S				
889	0	VV • /	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): Non-Null Count Dtype Column _____ ------ - -____ 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 5 714 non-null float64 Age 6 SibSp 891 non-null int64 7 int64 Parch 891 non-null 8 891 non-null object Ticket 9 Fare 891 non-null float64

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

204 non-null

889 non-null

memory usage: 83.7+ KB

In [10]:

10

11

test_df.describe

Cabin

Embarked

Out[10]:

<pre><bound \<="" method="" name="" ndframe.describe="" of="" passengerid="" pclass="" pre=""></bound></pre>										
0	`	892	3		Kelly, Mr. James					
1		893	3		Wilkes, Mrs. James (Ellen Needs)					
2		894	2		Myles, Mr. Thomas Francis					
3		895	3		Wirz, Mr. Albert					
4		896	3	Hirvon	en, Mrs. Alexander (Helga E Lindqvist)					
• •					•••					
413		1305	3		Spector, Mr. Woolf					
414		1306	1		Oliva y Ocana, Dona. Fermina					
415		1307	3		Saether, Mr. Simon Sivertsen					
416	1308 3 Ware, Mr. Frederick									
417		1309	3		Peter, Master. Michael J					
	_	_								
	Sex	Age	SibSp	Parch	Ticket Fare Cabin Embarked					
0	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 (315154 8.6625 NaN S					
3	male	27.0	0	0						
4	female	22.0	1	1	3101298 12.2875 NaN S					
• •	• • •	• • •	• • •	• • •	•••					
413	male	NaN	0	0	A.5. 3236 8.0500 NaN S					
414	female	39.0	0	0	PC 17758 108.9000 C105 (
415	male	38.5	0	0	SOTON/O.Q. 3101262 7.2500 NaN S					
416	male	NaN	0	0	359309 8.0500 NaN S					
417	male	NaN	1	1	2668 22.3583 NaN (

object

object

[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	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)

In [12]:

#to finding missing values

memory usage: 36.0+ KB

In [14]:

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

Out[14]:

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

In [15]:

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

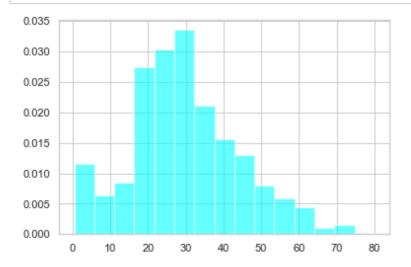
Out[15]:

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

utype. into-

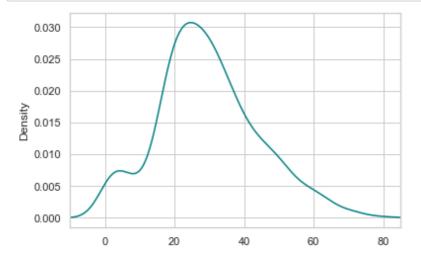
In [16]:

```
ax=train_df["Age"].hist(bins=15,density=True,stacked=True,color='cyan',alpha=0.6)
```



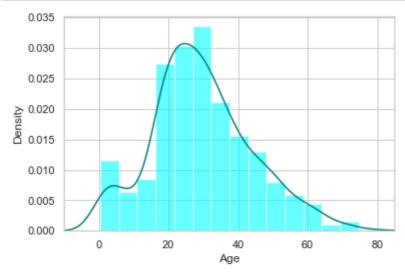
In [17]:

```
train_df["Age"].plot(kind='density',color='teal')
ax.set(xlabel='Age')
plt.xlim(-10,85)
plt.show()
```



In [18]:

```
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 [19]:

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

29.69911764705882

28.0

In [21]:

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

77.10437710437711

In [22]:

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

0.22446689113355783

In [23]:

```
\verb|print('Board passengers grouped by part of embartion(C=cherbourg,Q=Queenstown,S=Southmapton)|\\
```

Board passengers grouped by part of embartion(C=cherbourg,Q=Queenstown,S=South mapton):

In [24]:

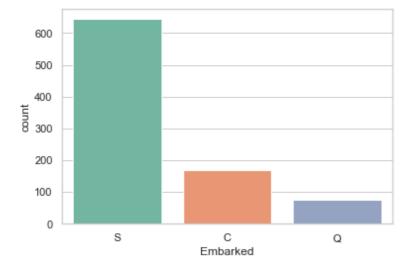
```
print(train_df['Embarked'].value_counts())
```

```
S 644
C 168
Q 77
```

Name: Embarked, dtype: int64

In [26]:

```
sns.countplot(x='Embarked',data=train_df,palette='Set2')
plt.show()
```



In [27]:

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

S

In [30]:

```
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)
```

In [31]:

```
train_data.drop('Cabin',axis=1,inplace=True)
```

In [32]:

```
train_data.isnull().sum()
```

Out[32]:

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 [33]:

train_data.head()

Out[33]:

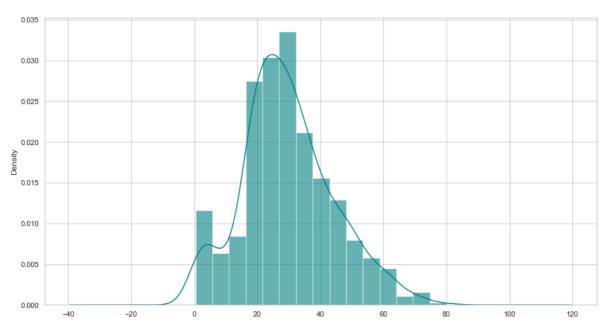
	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Emi
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	
3	4	1	1	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 [37]:

```
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')
```

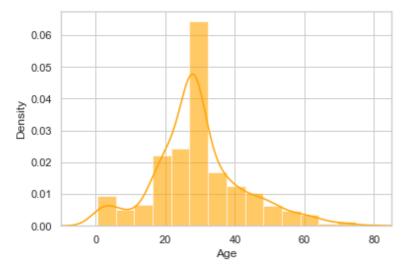
Out[37]:

<AxesSubplot:ylabel='Density'>



In [39]:

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



In [42]:

```
#create catagorical variable for travelling alone
train_data['TravelAlone']=np.where((train_data['SibSp']+train_data['Parch'])>0,0,1)
```

In [43]:

```
train_data.drop("SibSp",axis=1,inplace=True)
train_data.drop("Parch",axis=1,inplace=True)
```

In [45]:

```
#ctreate catagorical 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[45]:

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

In [46]:

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

Out[46]:

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

In [50]:

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

In [53]:

```
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)
```

In [54]:

```
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[54]:

	Age	Fare	TravelAlone	Pclass_1	Pclass_2	Pclass_3	Embarked_C	Embarked_Q	Embarke
0	34.5	7.8292	1	0	0	1	0	1	_
1	47.0	7.0000	0	0	0	1	0	0	
2	62.0	9.6875	1	0	1	0	0	1	
3	27.0	8.6625	1	0	0	1	0	0	
4	22.0	12.2875	0	0	0	1	0	0	
4									•

In [56]:

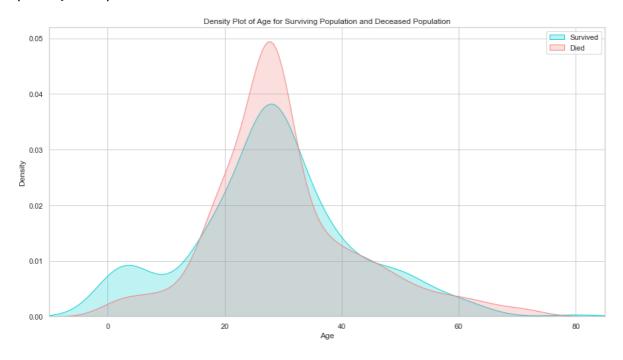
#EXPLORATORY DATA ANALYSIS

In [72]:

```
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 Deceased Population')
ax.set(xlabel='Age')
plt.xlim(-10,85)
```

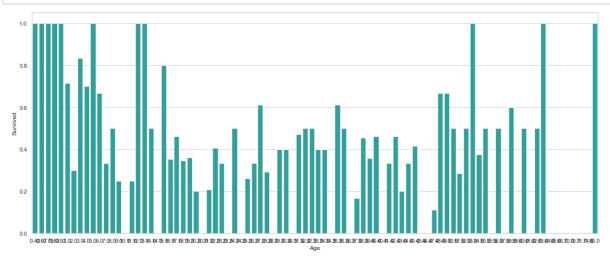
Out[72]:

(-10.0, 85.0)



In [73]:

```
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 [63]:

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

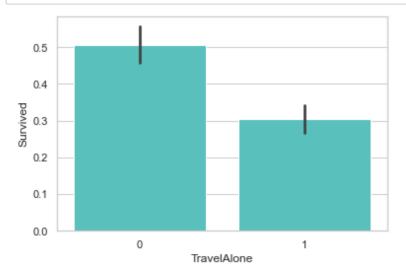
In [64]:

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

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

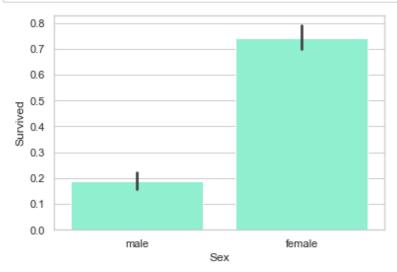
In [65]:

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



In [66]:

```
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 []: