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')
import warnings
warnings.simplefilter(action='ignore')
```

In [2]:

train_df=pd.read_csv(r"C:\Users\Krish\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	8.050
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\shaik\Downloads\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/Gender servival.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]:

train_df.shape

Out[7]:

(891, 12)

In [8]:

```
train_df.describe
```

Out[8]:

 0 1 2 3 4 886 887 888 889 890	nd meth	od NDFrame 1 2 3 4 5 887 888 889 890 891	.describe	e of 3 1 3 1 3 2 1 3 1 3 3	Passeng	erId	Surv	ived	Pclass	
n \						N	ame	Sex	Age	SibS
p \ 0				Braund,	Mr. Owe	n Har	ris	male	22.0	
1	Cuming	s, Mrs. Jo	hn Bradle	ey (Flore	nce Brig	gs Th	• • •	female	38.0	
2				Heikki	nen, Mis	s. La	ina	female	26.0	
0 3	F	utrelle, M	rs. Jacqı	ues Heath	(Lily M	ay Pe	el)	female	35.0	
1				Allen, M	r. Willi	am He	nry	male	35.0	
0							• • •		• • •	
886				Montv	ila, Rev	. Juo	zas	male	27.0	
0 887			Gral	nam, Miss	. Margar	et Ed	ith	female	19.0	
0 888		Johnsto	n, Miss.	Catherin	e Helen	"Carr	ie"	female	NaN	
1 889				Behr,	Mr. Kar	l How	ell	male	26.0	
0 890 0				Doo	ley, Mr.	Patr	ick	male	32.0	
0 1 2 3 4 886 887 888 889	Parch 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	P STON/O2.	Ticket 5 21171 C 17599 3101282 113803 373450 211536 112053 C. 6607 111369 370376	Fare 7.2500 71.2833 7.9250 53.1000 8.0500 13.0000 30.0000 23.4500 30.0000 7.7500	Cabin Eml NaN C85 NaN C123 NaN NaN B42 NaN C148 NaN		S C S S			

[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
	(1+(4/2	\ :+C4/E\ -L-	

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

memory usage: 83.7+ KB

In [10]:

test_df.describe

Out[10]:

<bou< th=""><th></th><th>d NDFr</th><th>ame.des</th><th>cribe of</th><th colspan="5">PassengerId Pclass</th></bou<>		d NDFr	ame.des	cribe of	PassengerId Pclass				
0	```	892	3			Kelly, M	lr. James		
1		893	3		Wilkes, Mrs. J				
2		894	2		=	Mr. Thomas	•		
3		895	3		•	Wirz, Mr	. Albert		
4		896	3	Hirvone	n, Mrs. Alexander (Helga E Li	ndqvist)		
 413		 1305	3			Spector, M	 Ir. Woolf		
414		1306	1		Oliva y Od	ana, Dona.			
415		1307	3		_	1r. Śimon S			
416		1308	3			lare, Mr. F			
417		1309	3			Master. M			
ed	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin Embark		
9 0	male	34.5	0	0	330911	7.8292	NaN		
Q 1	female	47.0	1	0	363272	7.0000	NaN		
S	Telliate	47.0	1	0	303272	7.0000	IVAIV		
2	male	62.0	0	0	240276	9.6875	NaN		
Q 3	male	27.0	0	0	315154	8.6625	NaN		
S									
4 S	female	22.0	1	1	3101298	12.2875	NaN		
• •			• • •	• • •	•••	• • •	•••		
 413	male	NaN	0	0	A.5. 3236	8.0500	NaN		
S					7.000 5250	0,000			
414 C	female	39.0	0	0	PC 17758	108.9000	C105		
415	male	38.5	0	0	SOTON/0.Q. 3101262	7.2500	NaN		
S	,		_	•	25222	0.0500			
416 S	male	NaN	0	0	359309	8.0500	NaN		
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	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
d+vn	os: float64/2) in+64(4) obj	oc+(5)

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

memory usage: 36.0+ KB

In [12]:

#to finding missing values

In [14]:

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

Out[14]:

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

dtype: int64

In [15]:

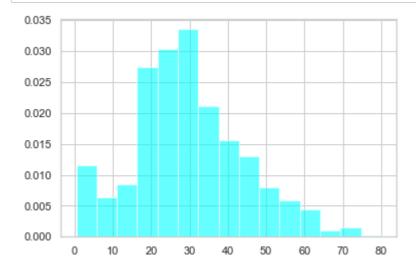
```
test_df.isnull().sum()
```

Out[15]:

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

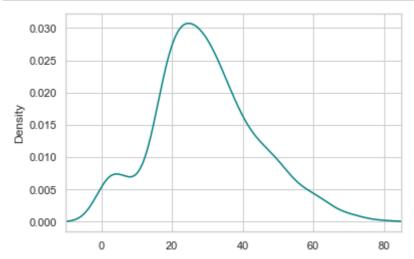
In [16]:

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



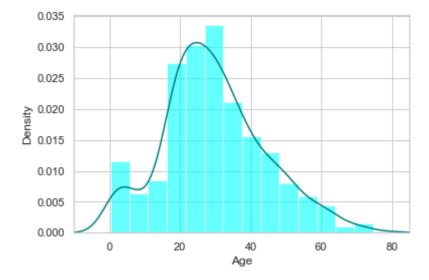
In [17]:

```
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]:

```
print('Board passengers grouped by part of embartion(C=cherbourg,Q=Queenstown,S=Southmap
```

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

In [24]:

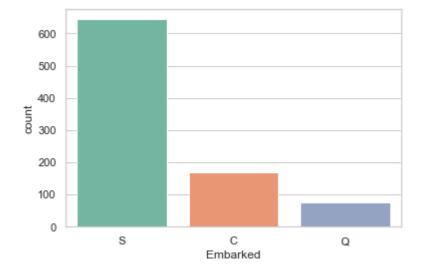
```
print(train_df['Embarked'].value_counts())
```

S 644 C 168 O 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]:

Pclass Name Sex Age SibSp Parch Ticket Fare Embarked	PassengerId	0
Name 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8	Survived	0
Sex & & & & & & & & & & & & & & & & & & &	Pclass	0
Age 6 SibSp 6 Parch 6 Ticket 6 Fare 6 Embarked 6	Name	0
SibSp 6 Parch 6 Ticket 6 Fare 6 Embarked 6	Sex	0
Parch 6 Ticket 6 Fare 6 Embarked 6	Age	0
Ticket 6 Fare 6 Embarked 6	SibSp	0
Fare 6 Embarked 6	Parch	0
Embarked 6	Ticket	0
	Fare	0
dtypo: int64	Embarked	0
utype. Into4	dtype: int64	

In [33]:

train_data.head()

Out[33]:

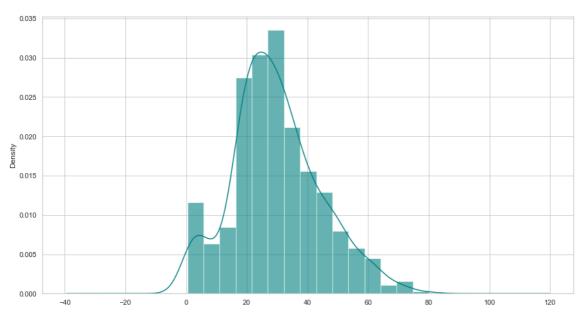
	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250
3	4	1	1	Futrelle,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

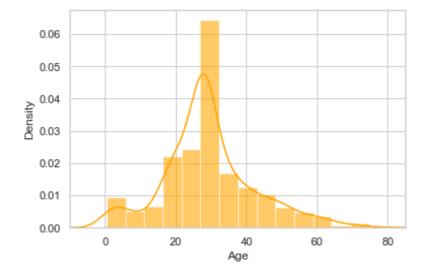
Out[37]:

<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	Embark
0	0	22.0	7.2500	0	0	0	1	0	
1	1	38.0	71.2833	0	1	0	0	1	
2	1	26.0	7.9250	1	0	0	1	0	
3	1	35.0	53.1000	0	1	0	0	0	
4	0	35.0	8.0500	1	0	0	1	0	
4									•

In [46]:

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

Out[46]:

PassengerId 0 Pclass 0 Name 0 Sex 0 86 Age SibSp 0 0 Parch 0 Ticket Fare 1 327 Cabin **Embarked** 0 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	Em
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]:

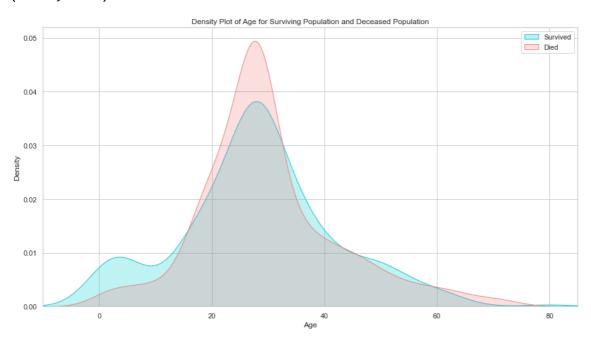
#EXPLORATORY DATA ANALYSIS

In [72]:

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

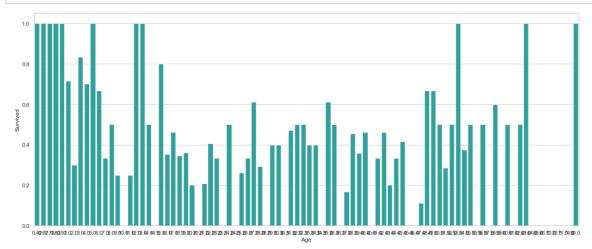
Out[72]:

(-10.0, 85.0)



In [73]:

```
plt.figure(figsize=(20,8))
avg_survival_byage = final_train[["Age", "Survived"]].groupby(['Age'], as_index=False).m
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
4
       0
886
       0
       0
887
       0
888
889
       0
890
       0
Name: IsMinor, Length: 891, dtype: int32
```

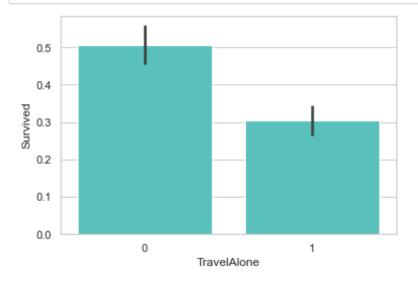
In [64]:

```
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
       0
415
416
       0
417
Name: IsMinor, Length: 418, dtype: int32
```

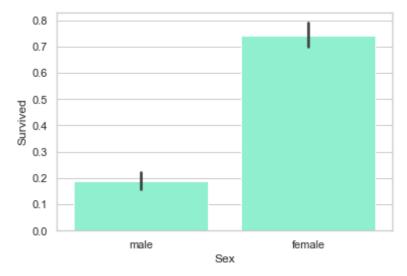
In [65]:

```
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 []: