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\RAMADEVI SURIPAKA\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
801 r	ows × 12 colu	ımne								
0911	UVV5 ^ 12 CUIU	6111118								

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In [3]:

train_df.head()

Out[3]:

	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 [4]:

train_df.tail()

Out[4]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	С
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	
4											•

In [5]:

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

Out[5]:

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cał
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 [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.tail()

Out[7]:

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cab
413	1305	3	Spector, Mr. Woolf	male	NaN	0	0	A.5. 3236	8.0500	Na
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	Na
416	1308	3	Ware, Mr. Frederick	male	NaN	0	0	359309	8.0500	Na
417	1309	3	Peter, Master. Michael J	male	NaN	1	1	2668	22.3583	Na
4										•

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
4							•

In [9]:

test_df.describe()

Out[9]:

	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 [10]:

train_df.shape

Out[10]:

(891, 12)

In [11]:

test_df.shape

Out[11]:

(418, 11)

In [12]:

```
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 5 714 non-null float64 Age 6 SibSp 891 non-null int64 7 int64 Parch 891 non-null 8 Ticket 891 non-null object 9 891 non-null float64 Fare 10 Cabin 204 non-null object 889 non-null 11 Embarked object

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

memory usage: 83.7+ KB

In [13]:

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
	63		

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

memory usage: 36.0+ KB

In [14]:

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

Out[14]:

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

In [15]:

test_df.isnull().sum()

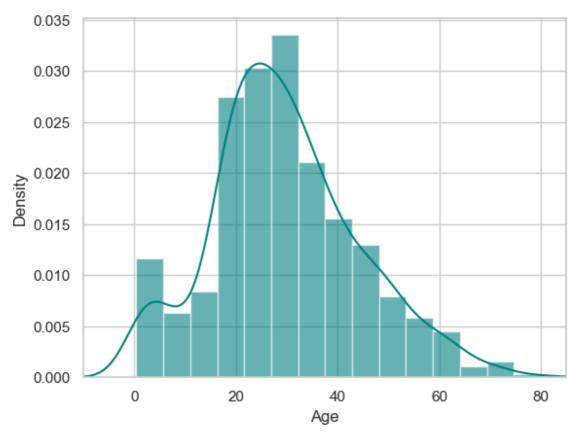
Out[15]:

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

dtype: int64

In [16]:

```
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.set(xlabel="Age")
plt.xlim(-10,85)
plt.show()
```



In [17]:

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

29.69911764705882

28.0

In [18]:

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

77.10437710437711

In [19]:

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

0.22446689113355783

In [21]:

```
print('Boarded passengers grouped by part of embarketion (C = Cherbourg,Q=Queenstown,S=S
print(train_df['Embarked'].value_counts())
sns.countplot(x='Embarked',data=train_df,palette='Set2')
plt.show()
```

Boarded passengers grouped by part of embarketion (C = Cherbourg,Q=Queenst own,S=South ampton):

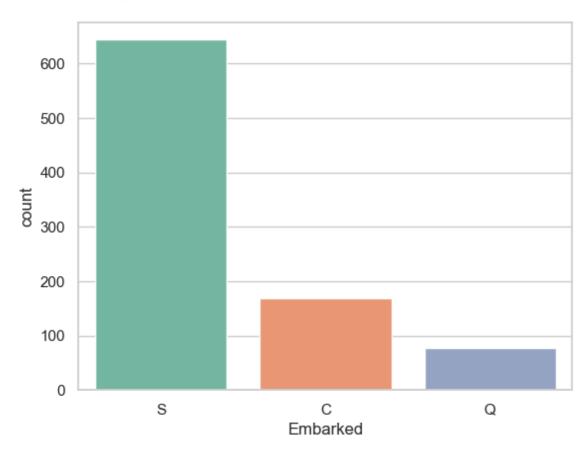
Embarked

S 644

C 168

Q 77

Name: count, dtype: int64



In [22]:

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

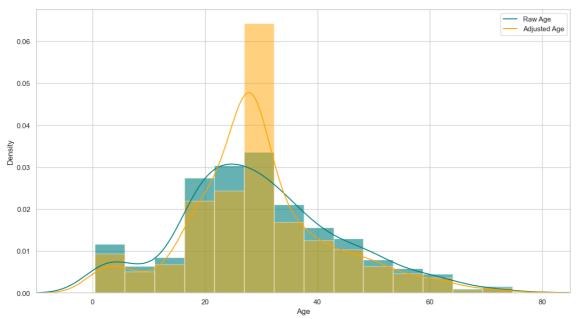
S

In [25]:

```
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 [26]:

```
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]:

```
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[27]:

	Survived	Age	SibSp	Parch	Fare	Pclass_1	Pclass_2	Pclass_3	Embarked_C	Emba
0	0	22.0	1	0	7.2500	False	False	True	False	
1	1	38.0	1	0	71.2833	True	False	False	True	
2	1	26.0	0	0	7.9250	False	False	True	False	
3	1	35.0	1	0	53.1000	True	False	False	False	
4	0	35.0	0	0	8.0500	False	False	True	False	
4										•

In [29]:

```
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(train_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()
```

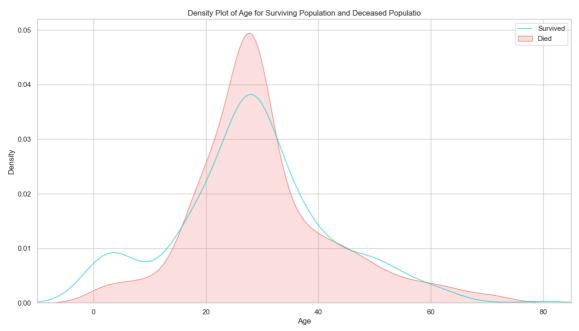
Out[29]:

	Survived	Age	SibSp	Parch	Fare	Pclass_1	Pclass_2	Pclass_3	Embarked_C	Emba
0	0	22.0	1	0	7.2500	False	False	True	False	
1	1	38.0	1	0	71.2833	True	False	False	True	
2	1	26.0	0	0	7.9250	False	False	True	False	
3	1	35.0	1	0	53.1000	True	False	False	False	
4	0	35.0	0	0	8.0500	False	False	True	False	
4										•

EXPLORATORY DATA ANALASYS

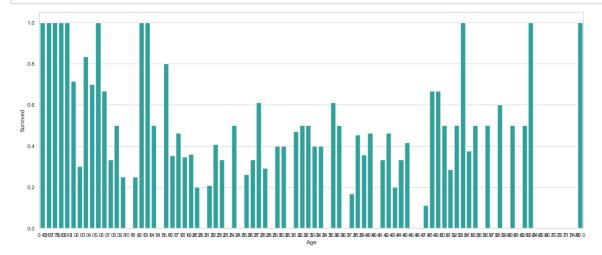
In [31]:

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



In [32]:

```
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 [33]:
```

```
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
       0
887
888
       0
       0
889
890
       0
Name: IsMinor, Length: 891, dtype: int32
In [34]:
final_test['IsMinor']=np.where(final_test['Age']<=16, 1, 0)</pre>
print(final_train['IsMinor'])
0
       0
1
       0
2
       0
3
       0
4
       0
       0
886
887
       0
       0
888
889
       0
890
Name: IsMinor, Length: 891, dtype: int32
In [ ]:
```