PROBLEM STATEMENT: To Predict And Analyze Which Gender Has A High Chance Of Survival At TheTime Of Disaster

IMPORT DATASETS, PYTHON PACKAGES AND LIBRARIES (02-06-2023)

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") #White background style for seaborn plots
sns.set(style="whitegrid", color_codes=True)
import warnings
warnings.simplefilter(action='ignore')

In [36]:

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

Out[36]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarke
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

localhost:8888/notebooks/Documents/GENDER.ipynb

In [4]:

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

Out[4]:

	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	male	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
413	1305	3	Spector, Mr. Woolf	male	NaN	0	0	A.5. 3236	8.0500	NaN	S
414	1306	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	108.9000	C105	С
415	1307	3	Saether, Mr. Simon Sivertsen	male	38.5	0	0	SOTON/O.Q. 3101262	7.2500	NaN	S
416	1308	3	Ware, Mr. Frederick	male	NaN	0	0	359309	8.0500	NaN	S
417	1309	3	Peter, Master. Michael J	male	NaN	1	1	2668	22.3583	NaN	С

418 rows × 11 columns

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	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	male	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 [7]: Ν

test_df.shape

Out[7]:

(418, 11)

In [8]: ▶

```
train_df.describe
```

Out[8]:

<box< td=""><td>nd meth</td><td>od NDFrame</td><td>.describe</td><td>e of</td><td>Passenger</td><td>Id Sur</td><td>vived I</td><td>class</td><td>\</td><td></td></box<>	nd meth	od NDFrame	.describe	e of	Passenger	Id Sur	vived I	class	\	
0		1	0	3						
1		2	1	1						
2		3	1	3						
3		4	1	1						
4		5	0	3						
• •		• • •	• • •	• • •						
886		887	0	2						
887		888	1	1						
888		889	0	3						
889		890	1	1						
890		891	0	3						
							_		c., c	
•						Name	Sex	_	SibSp	\
0					Mr. Owen		male	22.0	1	
1	Cuming	s, Mrs. Joh	nn Bradle				female	38.0	1	
2					nen, Miss.		female	26.0	0	
3	F	utrelle, M	rs. Jacqu				female		1	
4				Allen, M	lr. William	Henry	male	35.0	0	
••						- • • •	• • •	•••	• • •	
886					ila, Rev.		male	27.0	0	
887					. Margaret		female	19.0	0	
888		Johnstor	n, Miss.		e Helen "C			NaN	1	
889				-	Mr. Karl		male		0	
890				Doc	oley, Mr. P	atrick	male	32.0	0	
	Parch		Ticket	Eano	Cabin Emba	nkod				
0	0	Λ/1	5 21171	7.2500	NaN	S				
1	0		17599	71.2833	C85	C				
2										
	0	STON/02. 3		7.9250	NaN	S				
3	0		113803	53.1000	C123	S				
4	0		373450	8.0500	NaN	S				
 886			211536	13.0000	 NaN	 S				
887	0		112053	30.0000	B42	S				
888	2	lal //	112033 C. 6607	23.4500	NaN	S				
889	0	W./(111369	30.0000	NaN C148	C				
				7.7500						
890	0		370376	1.7500	NaN	Q				

[891 rows x 12 columns]>

In [9]:

C

Q

```
train_df.describe
```

Out[9]:

<box< td=""><td>nd meth</td><td>od NDFran</td><td>ne.describe</td><td>of</td><td>Passen</td><td>gerId</td><td>Survived</td><td>Pclass</td><td>\</td><td></td></box<>	nd meth	od NDFran	ne.describe	of	Passen	gerId	Survived	Pclass	\	
0		1	0	3	·					
1		2	1	1						
2		3	1	3						
3		4	1	1						
4		5	0	3						
		• • •								
886		887	0	2						
887		888	1	1						
888		889	0	3						
889		890	1	1						
890		891	0	3						
						Na	me Se	x Age	SibSp	\
0				Braund,	Mr. Ow	en Harr	is mal	e 22.0	1	
1	Cuming	s, Mrs. J	John Bradle	y (Flore	ence Bri	ggs Th.	femal	e 38.0	1	
2				Heikki	nen, Mi	ss. Lai	na femal	e 26.0	0	
3	F	utrelle,	Mrs. Jacqu	es Heath	n (Lily I	May Pee	l) femal	e 35.0	1	
4				Allen, N	۱r. Will:	iam Hen	ry mal	e 35.0	0	
• •										
886					/ila, Re [,]			e 27.0	0	
887				am, Miss				e 19.0	0	
888		Johnst	on, Miss.	Catherin	ne Helen	"Carri	e" femal	e NaN	1	
889				Behr,	Mr. Ka	rl Howe	ll mal	e 26.0	0	
890				Doo	oley, Mr	. Patri	.ck mal	e 32.0	0	
	Parch		Ticket	Fare	Cabin E	mbarked				
0	0	A	A/5 21171	7.2500	NaN	S	•			
1	0		PC 17599	71.2833	C85	C				
2	0	STON/02.	3101282	7.9250	NaN	S	,			
3	0		113803	53.1000	C123	S	1			
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	•			
000	_		444360	20 0000	C1 10	_				

111369 30.0000 C148

7.7500

NaN

370376

[891 rows x 12 columns]>

0

889

890

In [10]: M

```
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)
		7. KD	

memory usage: 83.7+ KB

In [11]:

test_df.describe

Out[11]:

<pre><bound \<="" method="" name="" ndframe.describe="" of="" passengerid="" pclass="" pre=""></bound></pre>								
0	\	892	3		Volly Mn Jam	0.5		
		-			Kelly, Mr. Jam			
1		893	3		Wilkes, Mrs. James (Ellen Need	,		
2		894	2		Myles, Mr. Thomas Franc			
3		895	3		Wirz, Mr. Albe			
4		896	3	Hirvon	en, Mrs. Alexander (Helga E Lindqvis	t)		
• •		• • •	• • •		•	• •		
413		1305	3		Spector, Mr. Woo	lf		
414		1306	1		Oliva y Ocana, Dona. Fermi	na		
415		1307	3		Saether, Mr. Simon Siverts	en		
416		1308	3		Ware, Mr. Frederi	ck		
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	Q		
1	female	47.0	1	0	363272 7.0000 NaN	S		
2	male	62.0	0	0	240276 9.6875 NaN	Q		
3	male	27.0	0	0	315154 8.6625 NaN	S		
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	C		
415	male	38.5	0	0	SOTON/0.0. 3101262 7.2500 NaN	S		
416	male	NaN	0	0	359309 8.0500 NaN	S		
417	male	NaN	1	1	2668 22.3583 NaN	C		
41/	шате	IVaIV	1	1	2000 22.3363 NdN	C		

[418 rows x 11 columns]>

```
H
In [12]:
test_df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 418 entries, 0 to 417
Data columns (total 11 columns):
 #
     Column
                  Non-Null Count Dtype
     PassengerId 418 non-null
 0
                                  int64
 1
                  418 non-null
     Pclass
                                  int64
 2
                  418 non-null
                                  object
     Name
 3
                  418 non-null
     Sex
                                  object
 4
                  332 non-null
                                  float64
     Age
 5
                                  int64
     SibSp
                  418 non-null
                                  int64
 6
     Parch
                  418 non-null
 7
     Ticket
                  418 non-null
                                  object
 8
     Fare
                  417 non-null
                                  float64
 9
     Cabin
                  91 non-null
                                  object
 10
     Embarked
                  418 non-null
                                  object
dtypes: float64(2), int64(4), object(5)
memory usage: 36.1+ KB
```

To Find Missing Values

```
M
In [13]:
train df.isnull().sum()
Out[13]:
PassengerId
Survived
Pclass
                  0
Name
                  0
Sex
                  0
               177
Age
                  0
SibSp
Parch
                  0
Ticket
Fare
                  0
Cabin
                687
Embarked
dtype: int64
                                                                                                           M
In [14]:
test_df.isnull().sum()
```

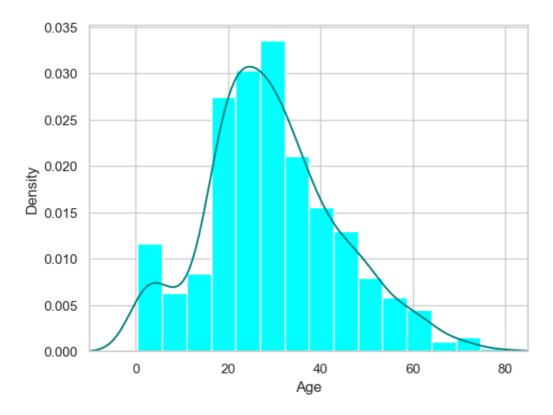
Out[14]:

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

localhost:8888/notebooks/Documents/GENDER.ipynb

```
In [15]:
```

```
ax = train_df["Age"].hist(bins=15, density=True, stacked=True, color='cyan')
train_df["Age"].plot(kind='density', color='teal')
ax.set(xlabel='Age')
plt.xlim(-10,85)
plt.show()
```



```
In [16]: ▶
```

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

29.69911764705882

28.0

```
In [17]: ▶
```

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

77.10437710437711

```
In [18]: ▶
```

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

0.22446689113355783

```
In [19]:

print('Boarded passengers grouped by port of embarkation (C = Cherbourg, Q = Queenstown,S =Southampton'
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 = Queenstown,S = South ampton

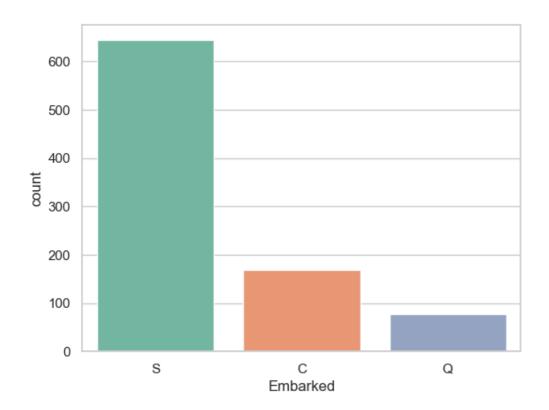
Embarked

S 644

C 168

Q 77

Name: count, dtype: int64



```
In [20]:
print(train_df['Embarked'].value_counts().idxmax())
```

S

```
In [21]:

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 [22]:

train_data.isnull().sum()

Out[22]:

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

In [23]:

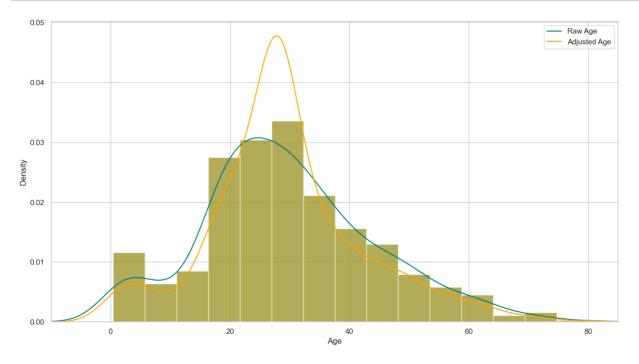
train_data.head()

Out[23]:

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

In [24]:

```
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_df["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 and drop some variables
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]:
```

```
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	Embarked_Q	Embarked_S
0	0	22.0	7.2500	1	False	False	True	False	False	True
1	1	38.0	71.2833	1	True	False	False	True	False	False
2	1	26.0	7.9250	1	False	False	True	False	False	True
3	1	35.0	53.1000	1	True	False	False	False	False	True
4	0	35.0	8.0500	1	False	False	True	False	False	True
4										•

```
M
In [27]:
test_df.isnull().sum()
Out[27]:
PassengerId
                    0
Pclass
                    0
Name
                    0
Sex
                   86
Age
                    0
SibSp
                    0
Parch
                    0
Ticket
                    1
Fare
Cabin
                  327
Embarked
                    0
dtype: int64
In [28]:
                                                                                                                           M
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('Name', axis=1, inplace=True)
testing.drop('Ticket', axis=1, inplace=True)
final_test = testing
final_test.head()
```

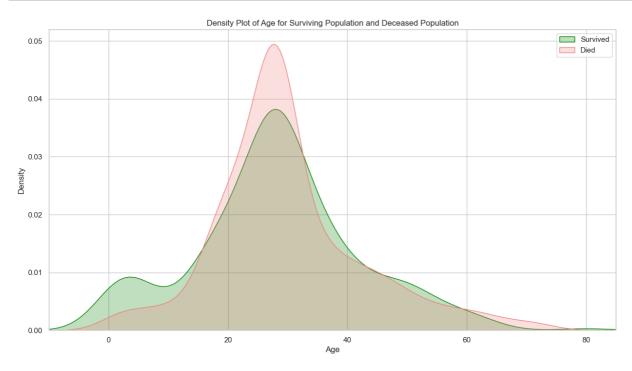
Out[28]:

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

EXPLORATORY DATA ANALYSIS

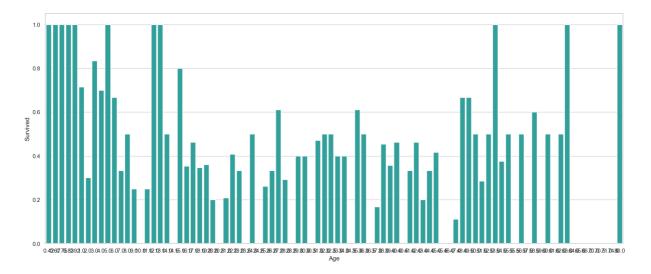
In [29]:

```
plt.figure(figsize=(15,8))
ax = sns.kdeplot(final_train["Age"][final_train.Survived == 1], color="green", shade=True)
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 [30]:
```

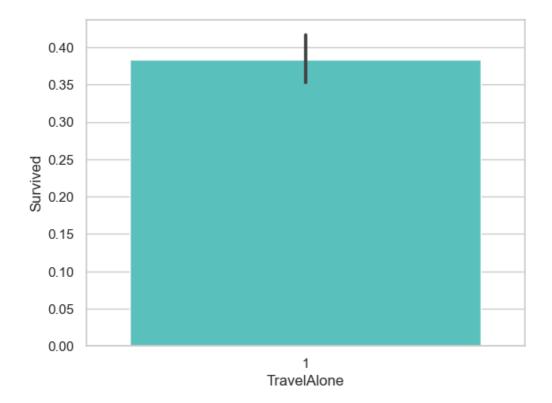
```
plt.figure(figsize=(20,8))
avg_survival_byage = final_train[["Age", "Survived"]].groupby(['Age'], as_index=False).mean()
g = sns.barplot(x='Age', y='Survived', data=avg_survival_byage, color="LightSeaGreen")
plt.show()z
```



```
H
In [31]:
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
887
       0
888
       0
889
       0
890
Name: IsMinor, Length: 891, dtype: int32
In [32]:
                                                                                                             H
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
415
       0
416
417
Name: IsMinor, Length: 418, dtype: int32
```

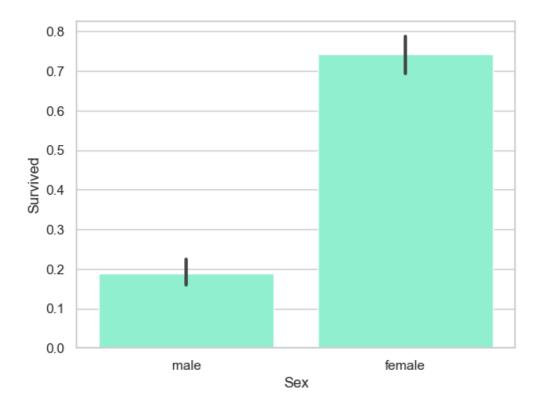
In [33]:

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



In [34]: ▶

```
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 []:	H