```
print('The features Matrix Has %d Rows and %d Columns(s)'%(features matrix.sha
In [21]:
         print('The Target Matrix has %d Rows and %d Columns(s)'%(np.array(target vector)
         The features Matrix Has 418 Rows and 2 Columns(s)
         The Target Matrix has 418 Rows and 1 Columns(s)
In [23]: | features_matrix_standardized=StandardScaler().fit_transform(features_matrix)
In [29]:
         algorithm=LogisticRegression(penalty='12',dual=False,tol=1e-4,C=1.0,fit_interd
In [30]:
         Logistic Regression Model=algorithm.fit(features matrix standardized,target ve
In [35]: observation=[[0.4,1]]
In [36]:
         predictions = Logistic Regression Model.predict(observation)
         print('The Model predicted The observation To Belong To class %s'%(predictions
         The Model predicted The observation To Belong To class [1]
         print('The Algorithm Was Trained To predict one of the two classes:%s'%(algorithm)
In [37]:
         The Algorithm Was Trained To predict one of the two classes:[0 1]
In [45]:
         print("""The Model says The probability of the observation We passed Belonging
         print()
         print("""The Model says The probability of the observation We passed Belonging
         The Model says The probability of the observation We passed Belonging To cla
         ss['0'] Is 0.0549611529834666
         The Model says The probability of the observation We passed Belonging To cla
         ss['1'] Is 0.0549611529834666
```

Data Analysis

To Predict and Analyse which gender has a High change of survival at the time of disaster.

Import datasets, python packages and libraries

```
In [1]: import numpy as np
   import pandas as pd
   from sklearn import preprocessing
   import matplotlib.pyplot as plt
   # plt.rc("font", size=14)
   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')
```

Out[2]:

	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
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	ma l e	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\91756\Documents\python\test.gender_submission.c
 test_df

Out[3]:

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin
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. A l bert	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	Ci
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	C
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	
4										1	

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	Embar
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.shape

Out[7]: (418, 11)

```
In [8]: train_df.describe
Out[8]: <bound method NDFrame.describe of
                                                    PassengerId
                                                                  Survived Pclass
                          1
                                     0
                                              3
                                                 \
         1
                          2
                                     1
                                              1
                          3
         2
                                     1
                                              3
         3
                          4
                                     1
                                              1
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                          5
                                     0
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                                   . . .
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         886
                       887
                                     0
                                              1
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                                     1
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                       889
                                     0
                                              3
                       890
                                              1
         889
                                     1
         890
                       891
                                     0
                                              3
                                                                                       SibSp
                                                                 Name
                                                                          Sex
                                                                                 Age
         0
                                           Braund, Mr. Owen Harris
                                                                                22.0
                                                                         male
                                                                                           1
         \
              Cumings, Mrs. John Bradley (Florence Briggs Th...
         1
                                                                       female
                                                                                38.0
                                                                                           1
         2
                                            Heikkinen, Miss. Laina
                                                                       female
                                                                                26.0
                                                                                           0
                    Futrelle, Mrs. Jacques Heath (Lily May Peel)
         3
                                                                       female
                                                                                35.0
                                                                                           1
         4
                                          Allen, Mr. William Henry
                                                                         male
                                                                                35.0
                                                                                           0
                                             Montvila, Rev. Juozas
                                                                                27.0
         886
                                                                         male
                                                                                           0
         887
                                      Graham, Miss. Margaret Edith
                                                                       female
                                                                                19.0
                                                                                           0
         888
                         Johnston, Miss. Catherine Helen "Carrie"
                                                                       female
                                                                                 NaN
                                                                                           1
         889
                                             Behr, Mr. Karl Howell
                                                                         male
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                                                                                           0
         890
                                                Dooley, Mr. Patrick
                                                                         male
                                                                                32.0
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              Parch
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                      STON/02. 3101282
                                           7.9250
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                                          13.0000
         887
                   0
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                                                     B42
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                                                                  S
         888
                   2
                             W./C. 6607
                                          23.4500
                                                     NaN
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                   0
         889
                                  111369
                                          30.0000
                                                    C148
         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):

#	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)

atypes: +10at64(2), int64(5), object

memory usage: 83.7+ KB

<pre>In [10]: test_df.describe</pre>

Out[10]:	<box< td=""><td>nd metho</td><td>od NDFr</td><td>ame.des</td><td>cribe of</td><td>PassengerId</td><td>Pclass</td><td></td><td></td></box<>	nd metho	od NDFr	ame.des	cribe of	PassengerId	Pclass					
	0		892	3			Kelly, N	۱r. Jan	ies \			
	1		893	3		Wilkes, Mrs. J						
	2		894	2		Myles, Mr. Thomas Francis						
	3		895	3		•	Wirz, Mr					
	4		896	3	Hirvone	en, Mrs. Alexander (
	• •							•	• •			
	413		1305	3			Spector, N	1r. Woo	olf			
	414		1306	1		Oliva y Od	ana, Dona.	. Fermi	.na			
	415		1307	3		Saether, M	lr. Simon S	Siverts	en			
	416		1308	3		W	lare, Mr. F	rederi	.ck			
	417		1309	3		Peter,	Master. M	ichae]	. 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			
	• •	• • •	• • •	• • •	• • •	• • •		• • •	• • •			
	41 3	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/O.Q. 3101262	7.2500	NaN	S			
	416	male	NaN	0	0	359309	8.0500	NaN	S			

1

2668

22.3583

NaN

[418 rows x 11 columns]>

NaN

male

417

C

```
In [11]: test_df.info()
```

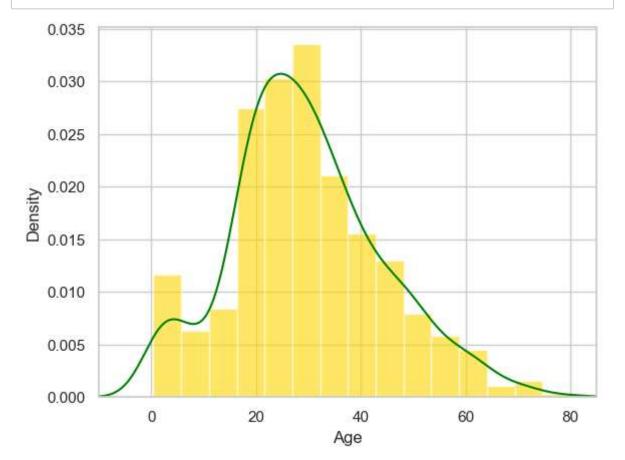
<class 'pandas.core.frame.DataFrame'> RangeIndex: 418 entries, 0 to 417 Data columns (total 11 columns):

#	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
dtvn	es: float64(2) $int64(4)$ ohi	ect(5)

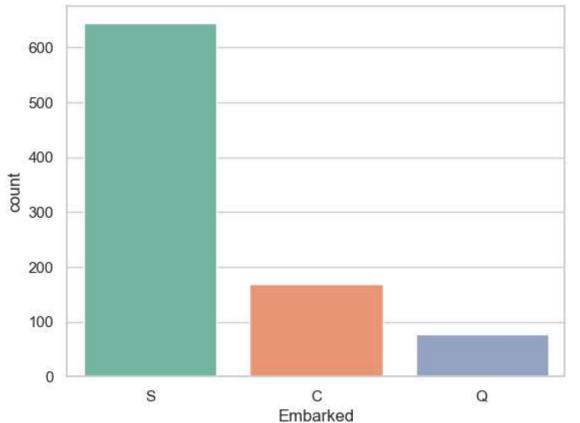
dtypes: float64(2), int64(4), object(5)

memory usage: 36.0+ KB

```
In [14]: | ax=train_df["Age"].hist(bins=15,density=True,stacked=True,color='gold',alpha=@)
         train_df["Age"].plot(kind='density',color='green')
         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
         print((train_df['Cabin'].isnull().sum()/train_df.shape[0])*100)
In [17]:
         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=Queenst
         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
         =Southampton):
         Embarked
              644
         C
              168
               77
         Name: count, dtype: int64
```



```
print(train df['Embarked'].value counts().idxmax())
         S
In [21]: train data = train df.copy()
In [22]:
         train_data['Age'].fillna(train_df["Age"].median(skipna=True),inplace=True)
         train data["Embarked"].fillna(train df['Embarked'].value counts().idxmax(),ing
         train data.drop('Cabin',axis=1,inplace=True)
In [23]: train_data.isnull().sum
Out[23]: <bound method NDFrame._add_numeric_operations.<locals>.sum of
                                                                             Passenger
         Id Survived Pclass
                                Name
                                                Age SibSp
                                         Sex
                                                            Parch
                                                                   Ticket
                    False
                               False
                                       False
                                              False False False
                                                                   False False
         0
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                                      False False False
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                                                                                  Fals
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                              False
                                      False
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               Fare
                     Embarked
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                        False
         3
              False
                        False
         4
              False
                        False
                . . .
                           . . .
         886
              False
                        False
              False
                        False
         887
         888
              False
                        False
         889
              False
                        False
         890 False
                        False
         [891 rows x 11 columns]>
```

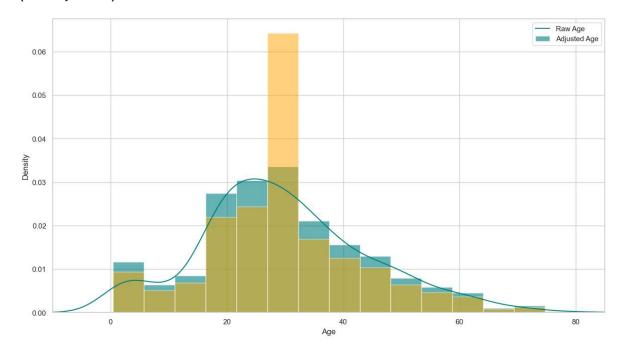
In [24]: train_data.head()

Out[24]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Eı
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 [27]: plt.figure(figsize= (15,8))
    ax = train_df["Age"].hist(bins=15,density=True,stacked=True,color='teal',alpha
    train_df["Age"].plot(kind='density',color='teal')
    ax =train_data["Age"].hist(bins=15,density=True,stacked=True,color='orange',alax.legend(['Raw Age','Adjusted Age'])
    ax.set(xlabel='Age')
    plt.xlim(-10,85)
```

Out[27]: (-10.0, 85.0)



```
In [38]: 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[38]:

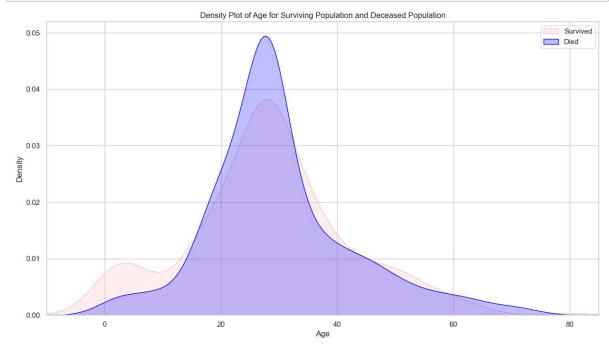
	Survived	Age	Fare	TravelAlone	Pclass_1	Pclass_2	Pclass_3	Embarked_C	Embarked_
0	0	22.0	7.2500	0	False	False	True	False	Fals
1	1	38.0	71.2833	0	True	False	False	True	Fals
2	1	26.0	7.9250	1	False	False	True	False	Fals
3	1	35.0	53.1000	0	True	False	False	False	Fals
4	0	35.0	8.0500	1	False	False	True	False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

```
In [39]: test df.isnull().sum()
Out[39]: PassengerId
                           0
         Pclass
                           0
         Name
                           0
         Sex
                           0
         Age
                          86
         SibSp
                           0
         Parch
                           0
         Ticket
         Fare
                           1
         Cabin
                         327
         Embarked
                           0
         dtype: int64
In [40]: |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, (
         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[40]:
```

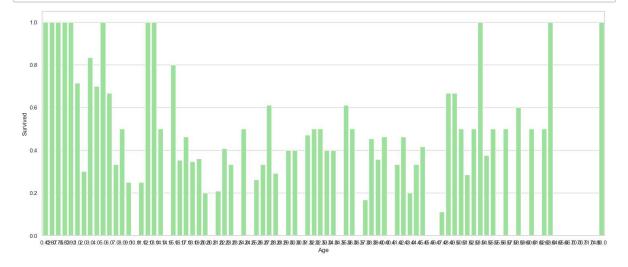
	Age	Fare	TravelAlone	Pclass_1	Pclass_2	Pclass_3	Embarked_C	Embarked_Q	Embarl
0	34.5	7.8292	1	False	False	True	False	True	
1	47.0	7.0000	0	False	False	True	False	False	
2	62.0	9.6875	1	False	True	False	False	True	
3	27.0	8.6625	1	False	False	True	False	False	
4	22.0	12.2875	0	False	False	True	False	False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

EXPLORATORY DATA ANALYSIS

In [58]: plt.figure(figsize=(15,8))
 ax=sns.kdeplot(final_train["Age"][final_train.Survived ==1],color="pink",shade
 sns.kdeplot(final_train["Age"][final_train.Survived == 0], color="blue", shade
 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()

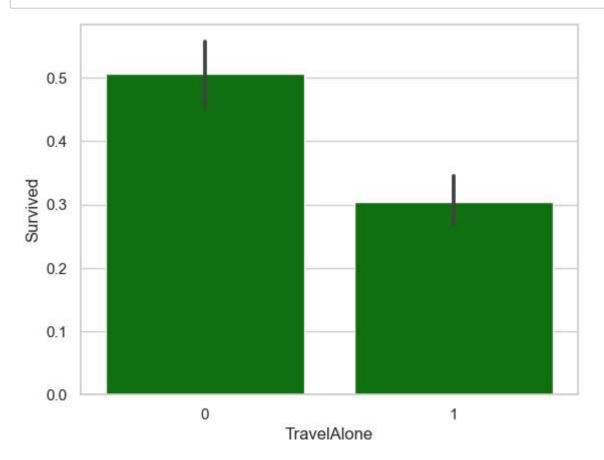


In [53]: plt.figure(figsize=(20,8))
 avg_survival_byage=final_train[["Age", "Survived"]].groupby(['Age'], as_index=
 g=sns.barplot(x='Age', y='Survived', data=avg_survival_byage, color="LightGree
 plt.show()

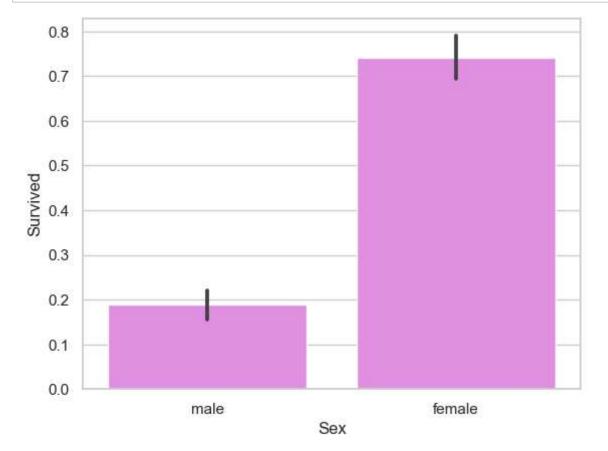


```
In [48]: | final_train['IsMinor']=np.where(final_train['Age']<=16, 1, 0)</pre>
          print(final_train['IsMinor'])
          0
                 0
          1
                 0
          2
                 0
          3
                 0
                 0
          886
                 0
          887
                 0
          888
                 0
          889
                 0
          890
          Name: IsMinor, Length: 891, dtype: int32
In [49]: 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
                 0
          417
          Name: IsMinor, Length: 418, dtype: int32
```

In [63]: sns.barplot(x='TravelAlone', y='Survived', data=final_train, color="Green")
plt.show()



```
In [59]: import seaborn as sns
import matplotlib.pyplot as plt
sns.barplot(x='Sex', y='Survived', data=train_df, color='violet')
plt.show()
```



In []: