Problem Statement:-

To predict and Analys which Gender has a high chance of survival at the time of disaster

Import datasets, python packages and libraries

Train Data

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]:

1 train_df=pd.read_csv(r"C:\Users\HP\OneDrive\Documents\train.gender_submission.csv")

2 train di

Out[2]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fi
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
891 r	ows × 12 colu	ımns								
4	12 3310									•

In [3]:

1 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]:

```
1 train_df.describe
```

Out[4]:

	_	LNSE		•	-		-	• • =	,	
	nd meth	od NDFrame			Passe	engerId	Surv	/ived P	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						
						N.	lama	Cov	A ~ o	c:h
Cn						IN	lame	Sex	Age	Sib
Sp				Pround	Mp (Ouan Han	nic	mala	22.0	
0 1 \				bi auiiu,	, MI (Owen Har	1.12	male	22.0	
1 \ 1	Cumina	s, Mrs. Jo	hn Dnadl	ov (Elono	onco Pi	niaac Th		fomalo	29 A	
1	Culliting	S, MI'S. JO	IIII DI AUT	ey (FION	ence bi	LISS2 III	• • •	тешате	38.0	
2				Hoikki	inon M	Miss. La	ina	female	26.0	
0				HETKKI		1133. La	ша	I CIII a T C	20.0	
3	E	utrelle, M	lns laco	uas Haatk	. (Lilv	y May Po	۱۱م	female	35.0	
1	'	uciciic, i	ii 3. Jacq	ucs ricaci	' (LII)	y May I C	C1)	TCIIIGIC	33.0	
4				Allen, M	n Wi ⁻	lliam He	nrv	male	35.0	
0				ALLEN, 1	w.	1114111 116	y	marc	33.0	
• •							•••	•••	•••	
886				Monty	/ila. F	Rev. Juo	zas	male	27.0	
0					,					
887			Gra	ham, Miss	. Mar	garet Ed	ith	female	19.0	
0				,	•	5				
888		Johnsto	n, Miss.	Catherin	ne Hele	en "Carr	ie"	female	NaN	
1			•							
889				Behr	, Mr. H	Karl How	ell	male	26.0	
0				•						
890				Doo	oley, N	Mr. Patr	ick	male	32.0	
0										
	Parch		Ticket	Fare	Cabin	Embarke	d			
0	0	A/	5 21171	7.2500	NaN		S			
1	0	P	C 17599	71.2833	C85		C			
2	0	STON/02.	3101282	7.9250	NaN		S			
3	0		113803	53.1000	C123		S			
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			
889	0		111369	30.0000	C148		C			
890	0		370376	7.7500	NaN		Q			
_			_							

[891 rows x 12 columns]>

```
In [5]:
```

```
1 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
4+	oc. £1oo+C4/2	\ in+64(E\ obi	o c + (F)

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

memory usage: 83.7+ KB

TO FIND MISSING VALUES

In [6]:

1 train_df.isnull().sum()

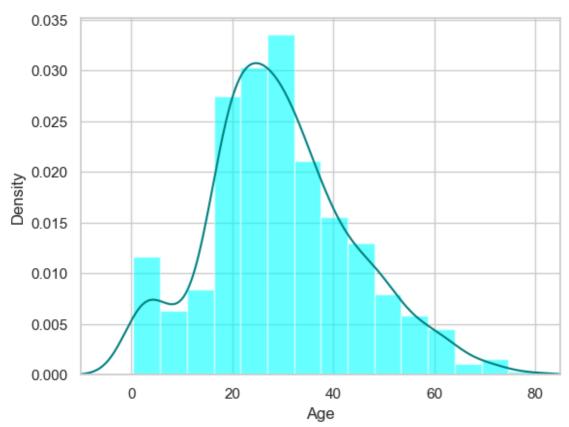
Out[6]:

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 [7]:

```
1 ax=train_df["Age"].hist(bins=15,density=True,stacked=True,color='cyan',alpha=0.6)
2 train_df['Age'].plot(kind='density',color='teal')
3 ax.set(xlabel='Age')
4 plt.xlim(-10,85)
5 plt.show()
```



In [8]:

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

29.69911764705882

28.0

In [9]:

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

77.10437710437711

In [10]:

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

0.22446689113355783

In [11]:

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

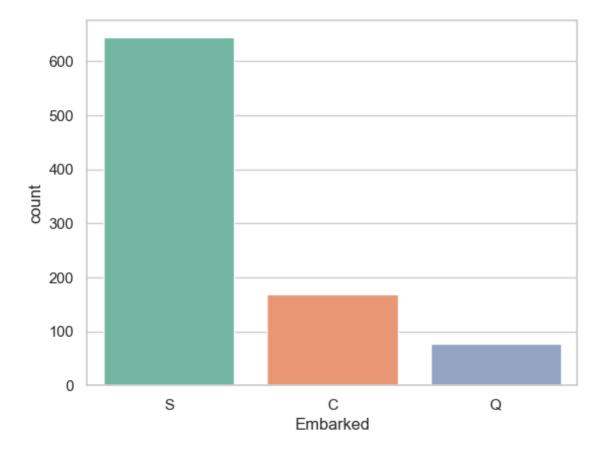
Boarded passengers grouped by port of embarked(C=Cherbourg,Q=Queenstown,S =Southampton):

Embarked

S 644

C 168

Name: count, dtype: int64



In [12]:

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

In [22]:

1 train_data.isnull().sum()

Out[22]:

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

In [23]:

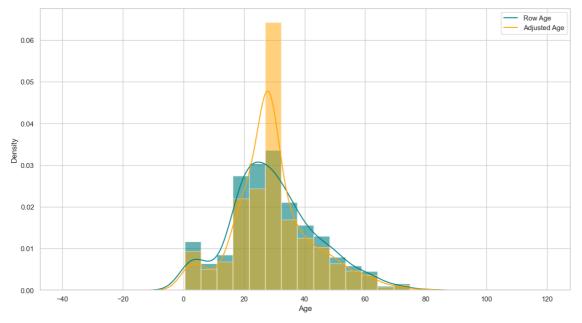
1 train_data.head()

Out[23]:

	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 [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')
df=train_data['Age'].hist(bins=15,density=True,stacked=True,color='orange',alpha=0.
train_data['Age'].plot(kind='density',color='orange')
ax.legend(['Row Age','Adjusted Age'])
ax.set(xlabel='Age')
plt.show()
```



In [25]:

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

Test data

In [27]:

```
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 [28]:

1 test_df=pd.read_csv(r"C:\Users\HP\OneDrive\Documents\test.gender_submission.csv")

2 test_df

Out[28]:

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Са
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	٨
2	894	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	٨
3	895	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	٨
4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	٨
413	1305	3	Spector, Mr. Woolf	male	NaN	0	0	A.5. 3236	8.0500	٨
414	1306	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	108.9000	С
415	1307	3	Saether, Mr. Simon Sivertsen	male	38.5	0	0	SOTON/O.Q. 3101262	7.2500	Ν
416	1308	3	Ware, Mr. Frederick	male	NaN	0	0	359309	8.0500	٨
417	1309	3	Peter, Master. Michael J	male	NaN	1	1	2668	22.3583	Ν

418 rows × 11 columns

In [29]:

1 test_df.head()

Out[29]:

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	En
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 [30]:

1 test_df.describe

Out[30]:

		d NDFr	ame.des	cribe of	PassengerId	Pclass		
Name 0		892	3			Kelly, M	Ar lam	iρς \
1		893	3		Wilkes, Mrs. J			
2		894	2			Mr. Thomas		
3		895	3		1.1,123,	Wirz, Mr		
4		896	3	Hirvone	n, Mrs. Alexander (-		
		• • •	•••		(8		••
413		1305	3			Spector, M	۱r. Woo	1f
414		1306	1		Oliva y Oc			
415		1307	3		Saether, M			
416		1308	3			lare, Mr. I		
417		1309	3			Master. N		
	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embar
ked								
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					2404000	40 0075		
4	female	22.0	1	1	3101298	12.2875	NaN	
S								
• •	• • •	• • •	• • •	• • •	• • •	• • •	• • •	
413	male	NaN	0	0	A.5. 3236	8.0500	NaN	
413 S	шате	IVAIV	V	V	A.5. 3230	8.0300	IVAIN	
3 414	female	39.0	0	0	PC 17758	108.9000	C105	
C C	Telliate	39.0	V	V	PC 17736	100.5000	CIOS	
415	male	38.5	0	0	SOTON/O.Q. 3101262	7.2500	NaN	
S S	mare	50.5	Ū	Ū	3010N/0.Q. 3101202	7.2300	IVAIV	
3 416	male	NaN	0	0	359309	8.0500	NaN	
5 S	mare	14014	0	Ü	555565	3.0500	NUN	
417	male	NaN	1	1	2668	22.3583	NaN	
C .			-	-	2000	22.3303		
_								

[418 rows x 11 columns]>

```
In [31]:
```

```
1 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 2 418 non-null object Name 3 418 non-null Sex object 4 332 non-null float64 Age 5 SibSp 418 non-null int64 6 418 non-null int64 Parch 7 Ticket 418 non-null object 8 float64 Fare 417 non-null

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

memory usage: 36.0+ KB

Cabin

10 Embarked

To Find Any Missing Values

91 non-null

418 non-null

In [33]:

9

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

object

object

Out[33]:

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

dtype: int64

In [36]:

```
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)
4 test_data.drop('Cabin',axis=1,inplace=True)
   test_data['TravelAlone']=np.where((test_data['SibSp']+test_data['Parch'])>0,0,1)
 5
   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)
11
   testing.drop('Ticket',axis=1,inplace=True)
12
13
14
   final_test=testing
   final_test.head()
15
```

Out[36]:

	Age	Fare	TravelAlone	Pclass_1	Pclass_2	Pclass_3	Embarked_C	Embarked_Q	Em
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									•

In [37]:

```
1 test_data.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 418 entries, 0 to 417
Data columns (total 9 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	418 non-null	float64
5	Ticket	418 non-null	object
6	Fare	418 non-null	float64
7	Embarked	418 non-null	object
8	TravelAlone	418 non-null	int32
عاد	C1+C4/2	\ :-+32/4\ :-+	C4(2) - I

dtypes: float64(2), int32(1), int64(2), object(4)

memory usage: 27.9+ KB

```
In [38]:
```

```
1 test_data.isnull().sum()
```

Out[38]:

PassengerId 0 Pclass 0 Name 0 Sex 0 0 Age Ticket 0 Fare 0 Embarked TravelAlone dtype: int64

In []:

1