```
In [1]: import pandas as pd
import numpy as np

In [2]: import matplotlib.pyplot as plt
import seaborn as sns

In [6]: #READING THE DATASET
    data = pd.read_csv("C:/Users/ashwa/Downloads/archive/train.csv")

In [7]: #CONVERTING TO DATAFRAME
    df=pd.DataFrame(data)
```

In [8]: #return first 5 rows of the dataframe
print(df.head)

0 1 2 3	nd metho	od NDFrame 1 2 3	0 1 1 1	3 1 3 1	ssengerId	Survive	d Pclas	s \	
4		5	0	3					
 886		 887	0	2					
887		888	1	1					
888		889	0	3					
889		890	1	1					
890		891	0	3					
						Name	Sex	Age	Sib
Sp	\					Name	Jex	Age	310
0				Braund,	Mr. Owe	n Harris	male	22.0	
1	Cumings	s, Mrs. Jo	hn Bradl	ey (Flore	ence Brig	gs Th	female	38.0	
1 2				Heikki	lnen, Mis	s. Laina	female	26.0	
0 3	Fı	utrelle, M	rs. Jacq	ues Heath	n (Lily M	ay Peel)	female	35.0	
1							-		
4				Allen, N	1r. Willi	am Henry	male	35.0	
0									
• •						• • •	• • •	• • •	
886 0				Mont	vila, Rev	. Juozas	male	27.0	
887 0			Gra	ham, Miss	. Margar	et Edith	female	19.0	
888 1		Johnsto	n, Miss.	Catherin	ne Helen	"Carrie"	female	NaN	
889				Behr,	Mr. Kar	l Howell	male	26.0	
0 890 0				Doo	oley, Mr.	Patrick	male	32.0	
	Parch		Ticket	Fare	Cabin Em	barked			
0	0	A/	5 21171	7.2500	NaN	S			
1	0	Р	C 17599	71.2833	C85	С			
2	0	STON/O2.				S			
3	0		113803	53.1000		S			
4	0		373450	8.0500	NaN	S			
••	• • •		 211526	12 0000	··· NaN	• • • •			
886 887	0		211536	13.0000		S S			
888 888	0 2	l.i /	112053 C. 6607	30.0000 23.4500		S S			
889	0	W•/	111369	30.0000		C			
890	0			7.7500	NaN	Q			
E001		12 1	-1.	500		¥			

[891 rows x 12 columns]>

In [9]: #Displays the no. of obs and features
df.shape

Out[9]: (891, 12)

0+1	[10]	
out	[TO]	•

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	(
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 [11]: 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)

memory usage: 83.7+ KB

```
In [13]: #will return the no. of missing records in each column
df.isnull().sum()
```

Out[13]: PassengerId Survived 0 Pclass 0 0 Name Sex 0 Age 177 SibSp 0 Parch 0 Ticket 0 0 Fare Cabin 687 Embarked 2

dtype: int64

In [14]: #fill missing value/null values with 0
df.fillna(0)

Out[14]:		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	0.0	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

3. Summary Statistics

In [15]: df.describe()

Out[15]:

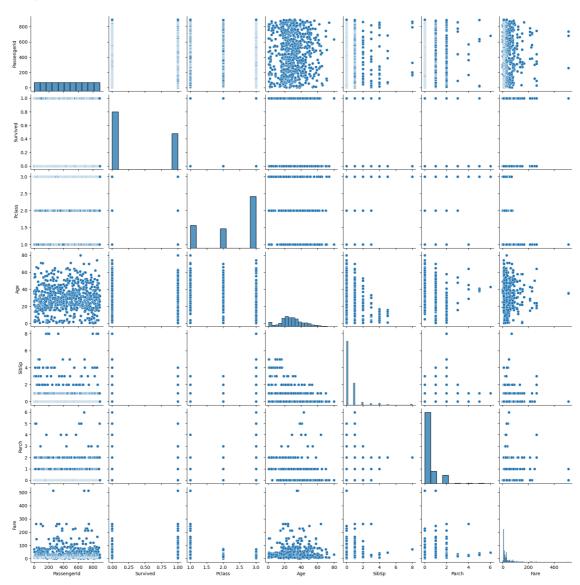
	Passengerld	Survived	Pclass	Age	SibSp	Parch	Fare
count	891.000000	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000
mean	446.000000	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208
std	257.353842	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429
min	1.000000	0.000000	1.000000	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							•

3.1. Pair Plot: Showing relationship between two categorical values

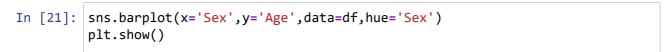
```
In [17]: # Check the column names in the DataFrame
print(data.columns)
```

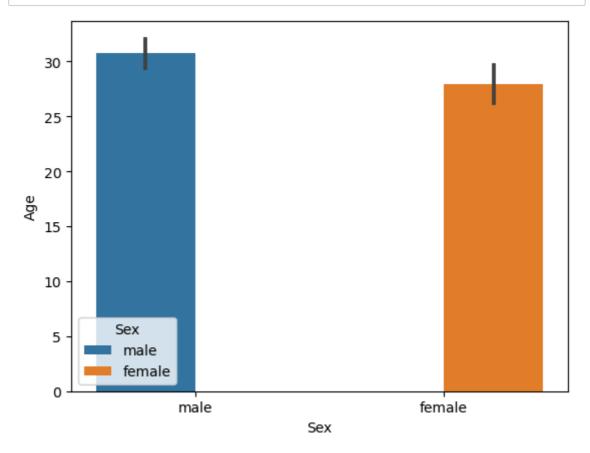
In [20]: # Dropping the specified columns and plotting the pairplot
 plt.figure(figsize=(13, 17))
 sns.pairplot(data=data.drop(['Sex', 'Embarked'], axis=1))
 plt.show()

<Figure size 1300x1700 with 0 Axes>



3.2 Bar Plot:showing relationship between categorical variables and continous variables





3.3. Heatmap: showing correlation between variables

In [23]: dfs = df.drop(['Name','Cabin','Sex','Ticket','Embarked'],axis=1)

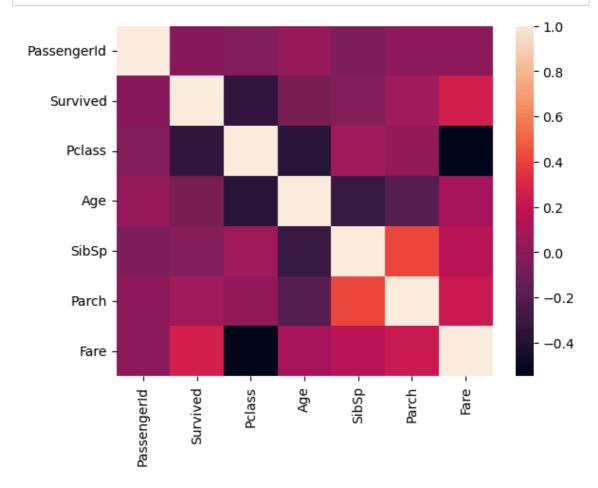
In [24]: dfs

Out[24]:

	Passengerld	Survived	Pclass	Age	SibSp	Parch	Fare
0	1	0	3	22.0	1	0	7.2500
1	2	1	1	38.0	1	0	71.2833
2	3	1	3	26.0	0	0	7.9250
3	4	1	1	35.0	1	0	53.1000
4	5	0	3	35.0	0	0	8.0500
886	887	0	2	27.0	0	0	13.0000
887	888	1	1	19.0	0	0	30.0000
888	889	0	3	NaN	1	2	23.4500
889	890	1	1	26.0	0	0	30.0000
890	891	0	3	32.0	0	0	7.7500

891 rows × 7 columns





In []: