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Importing the necessary libraries

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

Importing the dataset

```
In [2]: dataset = pd.read_csv("titanic_dataset.csv")
```

```
In [3]: dataset.head()
```

```
Out[3]:
```

	PassengerId	Name	Pclass	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	Braund, Mr. Owen Harris	3	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	Cumings, Mrs. John Bradley (Florence Briggs Th...	1	female	38.0	1	0	PC 17599	71.2833	C85	C
2	3	Heikkinen, Miss. Laina	3	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	Futrelle, Mrs. Jacques Heath (Lily May Peel)	1	female	35.0	1	0	113803	53.1000	C123	S
4	5	Allen, Mr. William Henry	3	male	35.0	0	0	373450	8.0500	NaN	S

```
In [4]: dataset.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   Name            891 non-null    object
2   Pclass          891 non-null    int64
3   Sex             891 non-null    object
4   Age            714 non-null    float64
5   SibSp           891 non-null    int64
6   Parch           891 non-null    int64
7   Ticket          891 non-null    object
8   Fare            891 non-null    float64
9   Cabin          204 non-null    object
10  Embarked        889 non-null    object
11  Survived        891 non-null    int64
dtypes: float64(2), int64(5), object(5)
memory usage: 83.7+ KB
```

```
In [5]: dataset.shape
```

```
Out[5]: (891, 12)
```

```
In [6]: dataset.describe
```

```

Out[6]: <bound method NDFrame.describe of PassengerId
Name Pclass \
0      1      Braund, Mr. Owen Harris      3
1      2 Cumings, Mrs. John Bradley (Florence Briggs Th... 1
2      3      Heikkinen, Miss. Laina      3
3      4 Futrelle, Mrs. Jacques Heath (Lily May Peel) 1
4      5      Allen, Mr. William Henry      3
..      ...      ...      ...
886     887      Montvila, Rev. Juozas      2
887     888      Graham, Miss. Margaret Edith      1
888     889      Johnston, Miss. Catherine Helen "Carrie" 3
889     890      Behr, Mr. Karl Howell      1
890     891      Dooley, Mr. Patrick      3

Sex Age SibSp Parch Ticket Fare Cabin Embarked \
0 male 22.0 1 0 A/5 21171 7.2500 NaN S
1 female 38.0 1 0 PC 17599 71.2833 C85 C
2 female 26.0 0 0 STON/O2. 3101282 7.9250 NaN S
3 female 35.0 1 0 113803 53.1000 C123 S
4 male 35.0 0 0 373450 8.0500 NaN S
.. ... ... ... ...
886 male 27.0 0 0 211536 13.0000 NaN S
887 female 19.0 0 0 112053 30.0000 B42 S
888 female NaN 1 2 W./C. 6607 23.4500 NaN S
889 male 26.0 0 0 111369 30.0000 C148 C
890 male 32.0 0 0 370376 7.7500 NaN Q

Survived
0 0
1 1
2 1
3 1
4 0
.. ...
886 0
887 1
888 0
889 1
890 0

```

[891 rows x 12 columns]>

Checking for null values

```
In [7]: dataset.isnull().any()
```

```
Out[7]: PassengerId    False
        Name          False
        Pclass        False
        Sex           False
        Age           True
        SibSp         False
        Parch         False
        Ticket        False
        Fare          False
        Cabin         True
        Embarked      True
        Survived      False
        dtype: bool
```

```
In [8]: dataset.isnull().sum()
```

```
Out[8]: PassengerId    0
        Name          0
        Pclass        0
        Sex           0
        Age          177
        SibSp         0
        Parch         0
        Ticket        0
        Fare          0
        Cabin        687
        Embarked      2
        Survived      0
        dtype: int64
```

```
In [9]: dataset.corr()
```

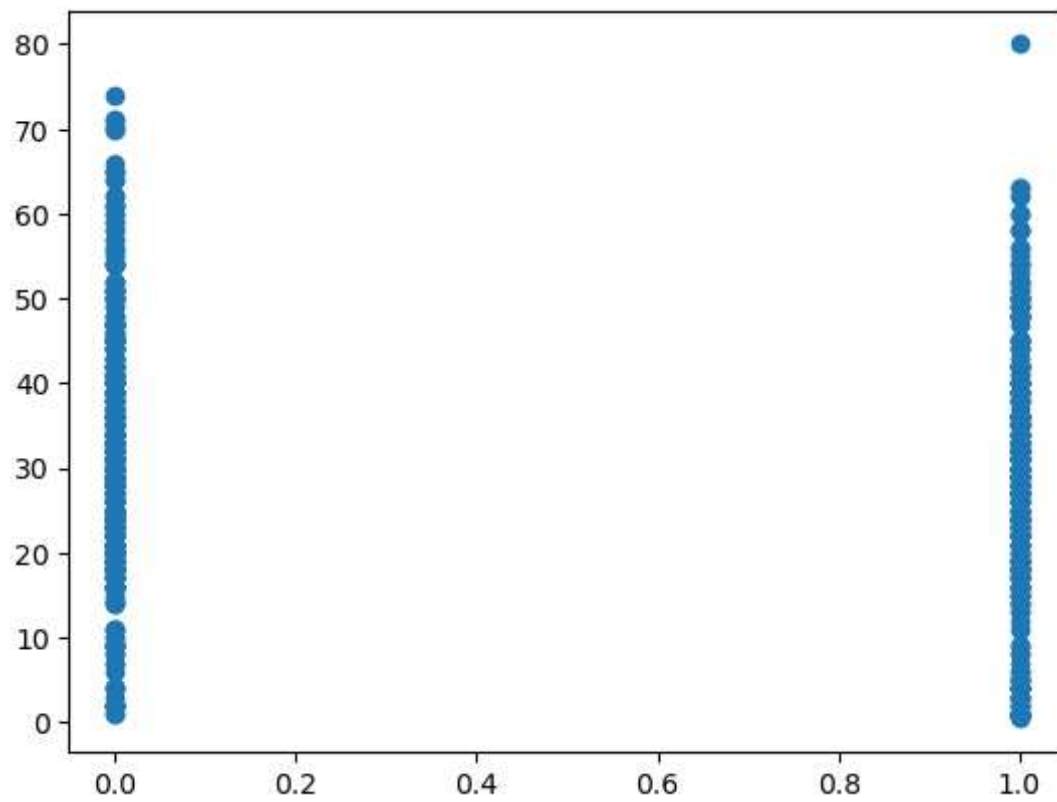
```
Out[9]:
```

	PassengerId	Pclass	Age	SibSp	Parch	Fare	Survived
PassengerId	1.000000	-0.035144	0.036847	-0.057527	-0.001652	0.012658	-0.005007
Pclass	-0.035144	1.000000	-0.369226	0.083081	0.018443	-0.549500	-0.338481
Age	0.036847	-0.369226	1.000000	-0.308247	-0.189119	0.096067	-0.077221
SibSp	-0.057527	0.083081	-0.308247	1.000000	0.414838	0.159651	-0.035322
Parch	-0.001652	0.018443	-0.189119	0.414838	1.000000	0.216225	0.081629
Fare	0.012658	-0.549500	0.096067	0.159651	0.216225	1.000000	0.257307
Survived	-0.005007	-0.338481	-0.077221	-0.035322	0.081629	0.257307	1.000000

Data Visualization

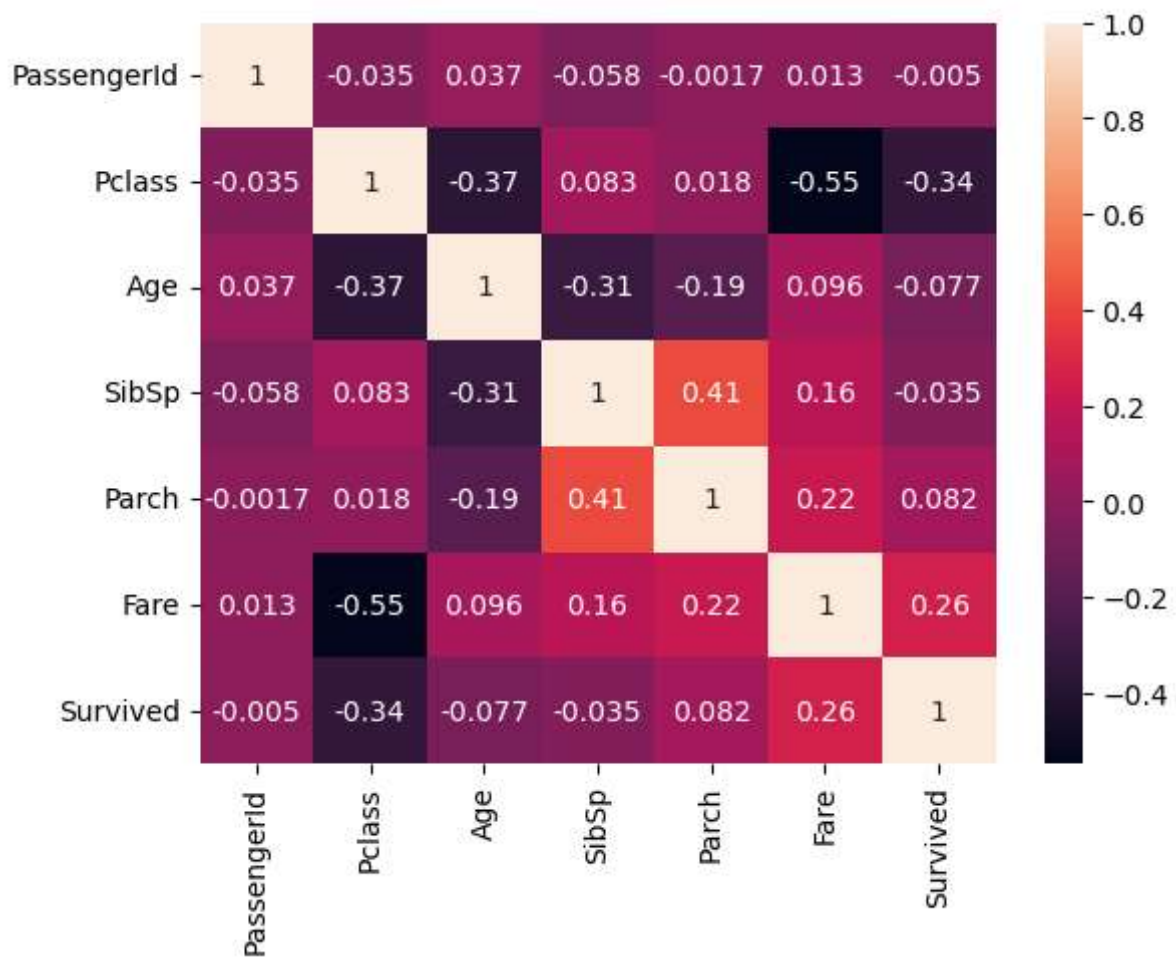
```
In [10]: plt.scatter(dataset["Survived"], dataset["Age"])
```

```
Out[10]: <matplotlib.collections.PathCollection at 0x1e2bfd66f90>
```



```
In [11]: sns.heatmap(dataset.corr(), annot = True)
```

```
Out[11]: <Axes: >
```



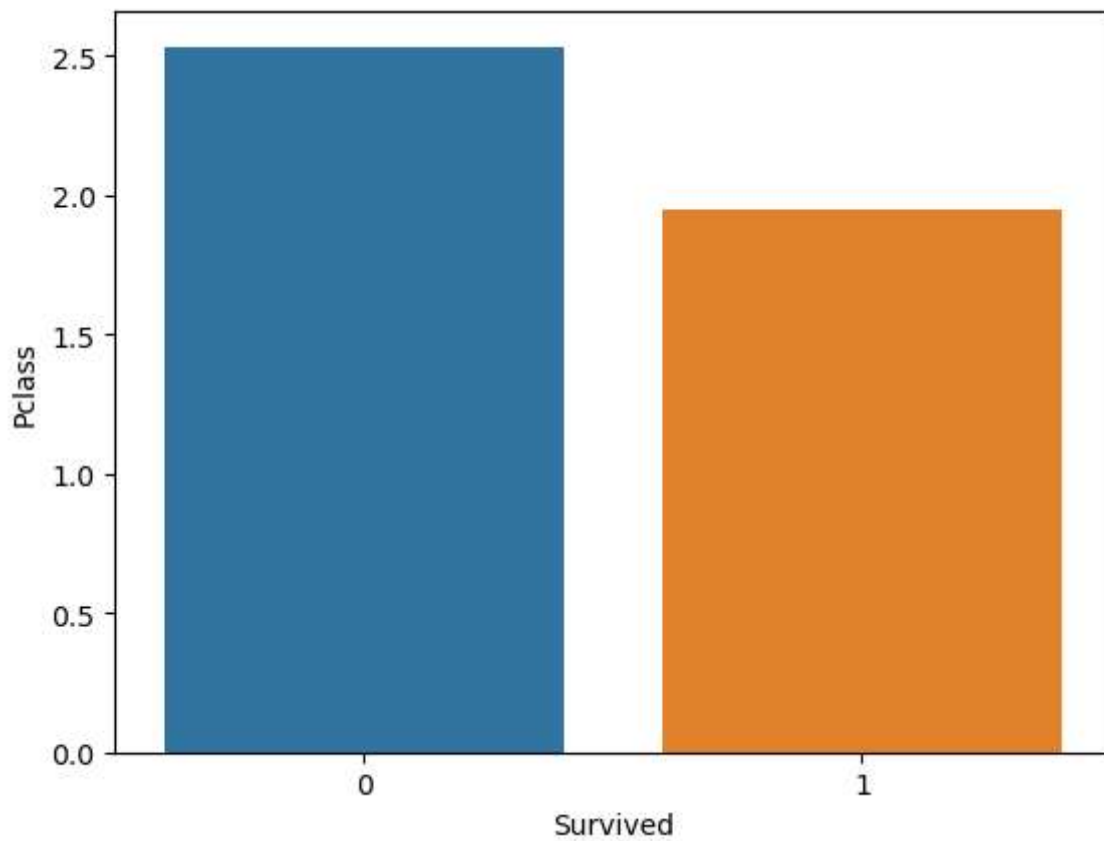
```
In [12]: sns.pairplot(dataset)
```

```
Out[12]: <seaborn.axisgrid.PairGrid at 0x1e2bfe1b490>
```



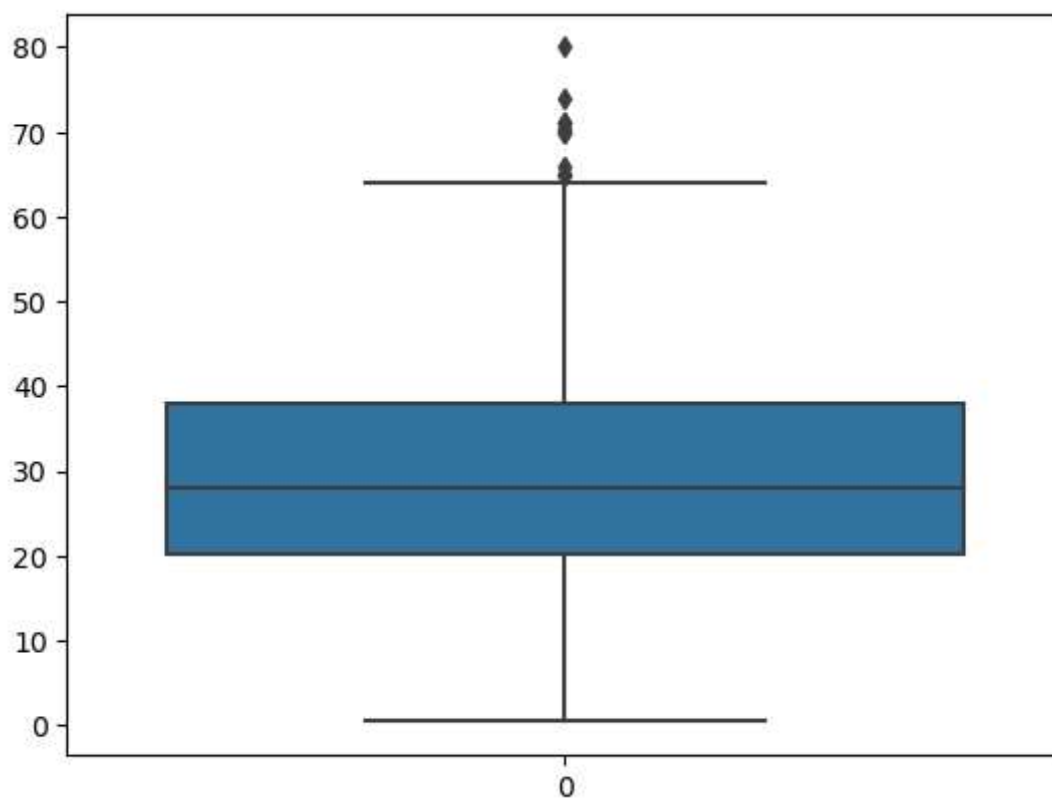
```
In [13]: sns.barplot(x = dataset["Survived"], y = dataset["Pclass"], ci = 0)
```

```
Out[13]: <Axes: xlabel='Survived', ylabel='Pclass'>
```



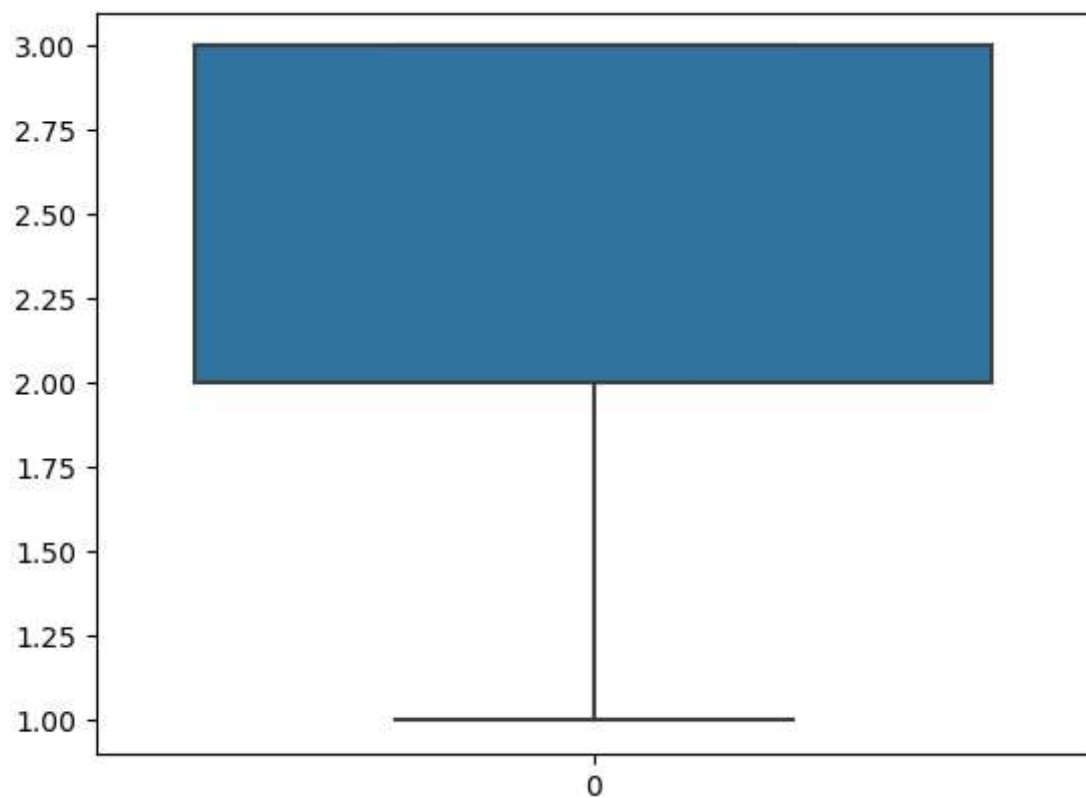
```
In [14]: sns.boxplot(dataset.Age)
```

```
Out[14]: <Axes: >
```



```
In [15]: sns.boxplot(dataset.Pclass)
```


Out[15]: <Axes: >



Splitting dependent and independent variables

In [16]: `dataset.head()`

Out[16]:

	PassengerId	Name	Pclass	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	Braund, Mr. Owen Harris	3	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	Cumings, Mrs. John Bradley (Florence Briggs Th...)	1	female	38.0	1	0	PC 17599	71.2833	C85	C
2	3	Heikkinen, Miss. Laina	3	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	Futrelle, Mrs. Jacques Heath (Lily May Peel)	1	female	35.0	1	0	113803	53.1000	C123	S
4	5	Allen, Mr. William Henry	3	male	35.0	0	0	373450	8.0500	NaN	S

In [17]: `x = dataset.drop(columns = ["Survived", "PassengerId", "Name", "Ticket", "Cabin"])`

In [18]: `x`

Out[18]:

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	male	22.0	1	0	7.2500	S
1	1	female	38.0	1	0	71.2833	C
2	3	female	26.0	0	0	7.9250	S
3	1	female	35.0	1	0	53.1000	S
4	3	male	35.0	0	0	8.0500	S
...
886	2	male	27.0	0	0	13.0000	S
887	1	female	19.0	0	0	30.0000	S
888	3	female	NaN	1	2	23.4500	S
889	1	male	26.0	0	0	30.0000	C
890	3	male	32.0	0	0	7.7500	Q

891 rows × 7 columns

In [19]: `x.shape`

Out[19]: (891, 7)

In [20]: `type(x)`

Out[20]: `pandas.core.frame.DataFrame`

In [21]: `y = dataset["Survived"]`

In [22]: `y.head()`

Out[22]:

0	0
1	1
2	1
3	1
4	0

Name: Survived, dtype: int64

In [23]: `type(y)`

Out[23]: `pandas.core.series.Series`

Encoding

In [24]: `x.head()`

Out[24]:

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	male	22.0	1	0	7.2500	S
1	1	female	38.0	1	0	71.2833	C
2	3	female	26.0	0	0	7.9250	S
3	1	female	35.0	1	0	53.1000	S
4	3	male	35.0	0	0	8.0500	S

In [25]: `from sklearn.preprocessing import LabelEncoder`
`le = LabelEncoder()`

In [26]: `x["Sex"] = le.fit_transform(x["Sex"])`

In [27]: `x.head()`

Out[27]:

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	1	22.0	1	0	7.2500	S
1	1	0	38.0	1	0	71.2833	C
2	3	0	26.0	0	0	7.9250	S
3	1	0	35.0	1	0	53.1000	S
4	3	1	35.0	0	0	8.0500	S

```
In [28]: print(le.classes_)
```

```
['female' 'male']
```

```
In [29]: mapping = dict(zip(le.classes_, range(len(le.classes_))))  
mapping
```

```
Out[29]: {'female': 0, 'male': 1}
```

```
In [30]: x["Embarked"] = le.fit_transform(x["Embarked"])
```

```
In [31]: x.head()
```

```
Out[31]:
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	1	22.0	1	0	7.2500	2
1	1	0	38.0	1	0	71.2833	0
2	3	0	26.0	0	0	7.9250	2
3	1	0	35.0	1	0	53.1000	2
4	3	1	35.0	0	0	8.0500	2

```
In [32]: print(le.classes_)
```

```
['C' 'Q' 'S' nan]
```

```
In [33]: mapping = dict(zip(le.classes_, range(len(le.classes_))))  
mapping
```

```
Out[33]: {'C': 0, 'Q': 1, 'S': 2, nan: 3}
```

```
In [34]: x.head()
```

```
Out[34]:
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	1	22.0	1	0	7.2500	2
1	1	0	38.0	1	0	71.2833	0
2	3	0	26.0	0	0	7.9250	2
3	1	0	35.0	1	0	53.1000	2
4	3	1	35.0	0	0	8.0500	2

Feature Scaling

```
In [35]: from sklearn.preprocessing import MinMaxScaler  
ms = MinMaxScaler()
```

```
In [36]: x_Scaled = pd.DataFrame(ms.fit_transform(x), columns=x.columns)
```

```
In [37]: x_Scaled.head()
```

```
Out[37]:
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	1.0	1.0	0.271174	0.125	0.0	0.014151	0.666667
1	0.0	0.0	0.472229	0.125	0.0	0.139136	0.000000
2	1.0	0.0	0.321438	0.000	0.0	0.015469	0.666667
3	0.0	0.0	0.434531	0.125	0.0	0.103644	0.666667
4	1.0	1.0	0.434531	0.000	0.0	0.015713	0.666667

Splitting data into training and testing

```
In [38]: from sklearn.model_selection import train_test_split  
X_train, X_test, y_train, y_test = train_test_split(x_Scaled, y, test_size =0.2, random
```

```
In [39]: print(X_train.shape, X_test.shape, y_train.shape, y_test.shape)  
  
(712, 7) (179, 7) (712,) (179,)
```

```
In [ ]:
```