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- REG NO: 21BCE7773
- CAMPUS: VIT-AP
- Assignment 3 on sept 15
- Morning Slot (10-12 am)
- Google colab Link: <https://colab.research.google.com/drive/119xWf81VtgwJAjWpFJCz9J4gsBqF0cQP?usp=sharing>

```
# Data Preprocessing.
# Import the Libraries.
# Import the dataset
# Checking for Null Values.

# Data Visualization.
# Outlier Detection
# Splitting Dependent and Independent variables
# Encoding
# Feature Scaling.
# Splitting Data into Train and Test.
```

+ Code

+ Text

Import the Libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

from google.colab import files
uploaded = files.upload()



Choose Files

 Titanic-Dataset.csv
• Titanic-Dataset.csv(text/csv) - 61194 bytes, last modified: 9/20/2023 - 100% done
Saving Titanic-Dataset.csv to Titanic-Dataset.csv
```

```
df = pd.read_csv("/content/Titanic-Dataset.csv")
```

```
df.head()
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin
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

```
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
```

```
dtypes: float64(2), int64(5), object(5)
memory usage: 83.7+ KB

df.describe()
```

	PassengerId	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

```
df.corr()
```

<ipython-input-8-2f6f6606aa2c>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, df.corr()

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

```
df.corr().Survived.sort_values(ascending = False)
```

<ipython-input-9-936bc0a2ea37>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, df.corr().Survived.sort\_values(ascending = False)

```
Survived      1.000000
Fare          0.257307
Parch         0.081629
PassengerId   -0.005007
SibSp         -0.035322
Age           -0.077221
Pclass       -0.338481
Name: Survived, dtype: float64
```

Handling Missing/Null Values

```
df.isnull().any()
```

PassengerId	False
Survived	False
Pclass	False
Name	False
Sex	False
Age	True
SibSp	False
Parch	False
Ticket	False
Fare	False
Cabin	True
Embarked	True

```
dtype: bool

sum(df.Cabin.isnull())
```

```
687

sum(df.Age.isnull())

177

df["Age"].fillna(df["Age"].mean(),inplace=True)

sum(df.Embarked.isnull())

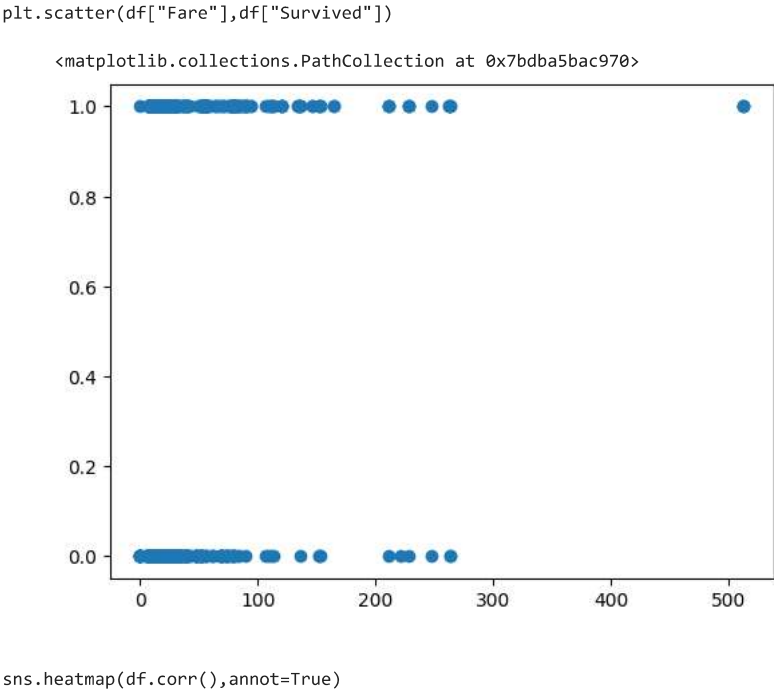
2

df["Embarked"].fillna(df["Embarked"].mode()[0],inplace=True)

df.describe()
```

	PassengerId	Survived	Pclass	Age	SibSp	Parch	Fare	
count	891.000000	891.000000	891.000000	891.000000	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	13.002015	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	22.000000	0.000000	0.000000	7.910400	
50%	446.000000	0.000000	3.000000	29.699118	0.000000	0.000000	14.454200	
75%	668.500000	1.000000	3.000000	35.000000	1.000000	0.000000	31.000000	
max	891.000000	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200	

Data Visualization

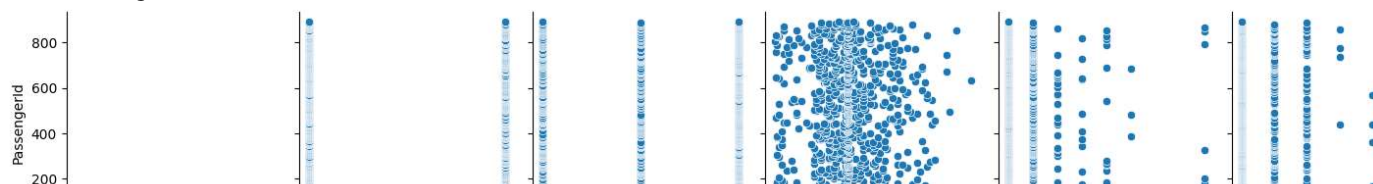


```
<ipython-input-18-8df7bcac526d>:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future versior
sns.heatmap(df.corr(),annot=True)
<Axes: >
```



```
sns.pairplot(df)
```

```
<seaborn.axisgrid.PairGrid at 0x7bdba39d7af0>
```



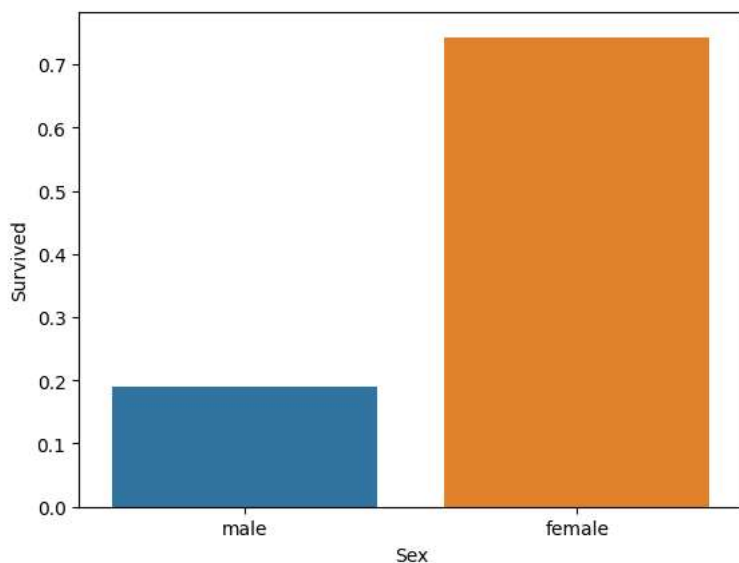
```
sns.barplot(x=df["Sex"],y=df["Survived"],ci=0)
```

```
<ipython-input-20-8ae461271d98>:1: FutureWarning:
```

The `ci` parameter is deprecated. Use `errorbar=('ci', 0)` for the same effect.

```
sns.barplot(x=df["Sex"],y=df["Survived"],ci=0)
```

```
<Axes: xlabel='Sex', ylabel='Survived'>
```



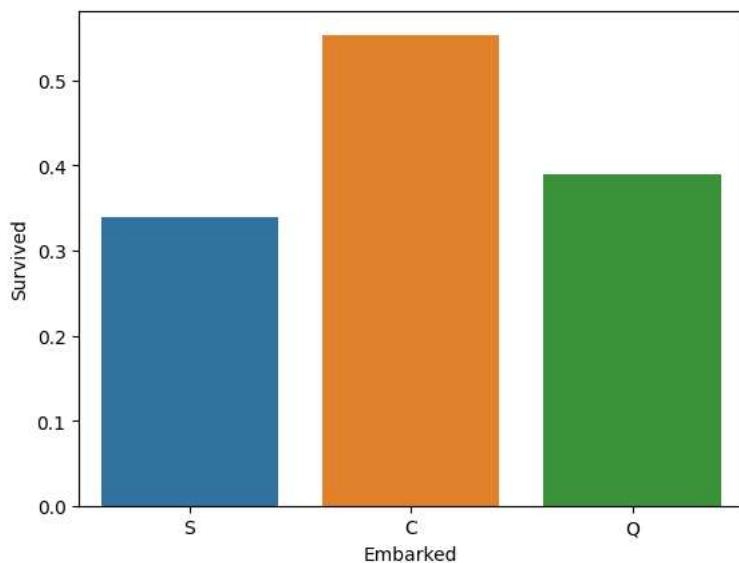
```
sns.barplot(x=df["Embarked"],y=df["Survived"],ci=0)
```

```
<ipython-input-21-d5b0276940a6>:1: FutureWarning:
```

The `ci` parameter is deprecated. Use `errorbar=('ci', 0)` for the same effect.

```
sns.barplot(x=df["Embarked"],y=df["Survived"],ci=0)
```

```
<Axes: xlabel='Embarked', ylabel='Survived'>
```

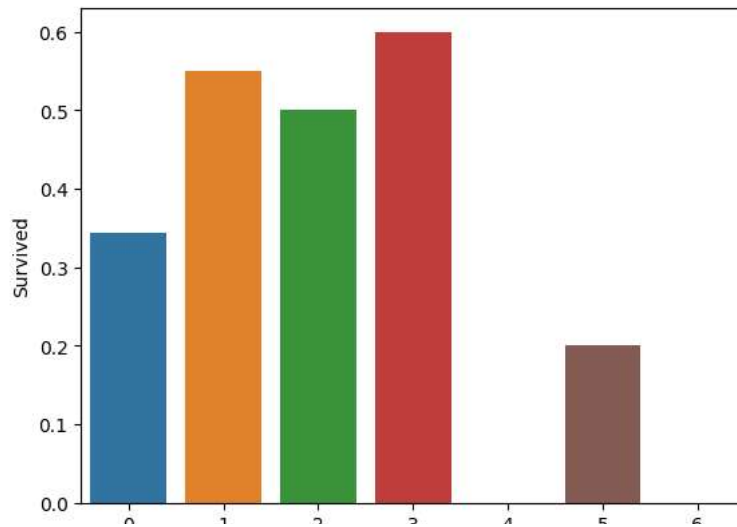


```
sns.barplot(x=df["Parch"],y=df["Survived"],ci=0)
```

```
<ipython-input-22-a1496feaf8>:1: FutureWarning:
```

The `ci` parameter is deprecated. Use `errorbar=('ci', 0)` for the same effect.

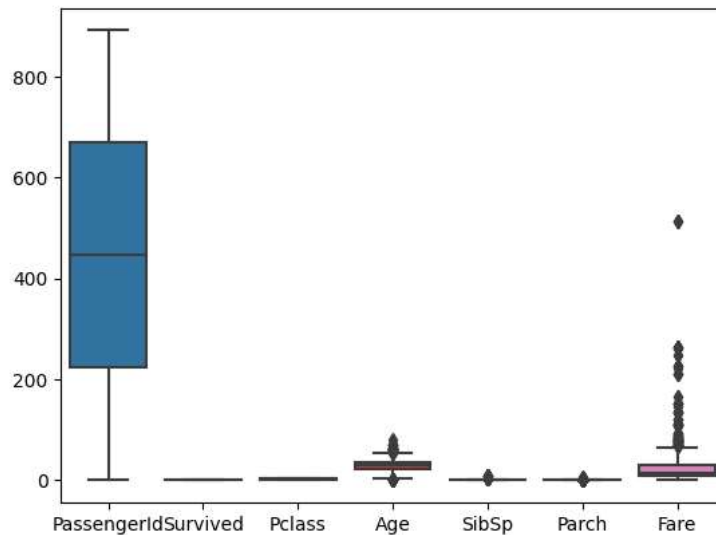
```
sns.barplot(x=df["Parch"],y=df["Survived"],ci=0)
<Axes: xlabel='Parch', ylabel='Survived'>
```



### Outlier Detection

```
sns.boxplot(df)
```

<Axes: >



```
sns.boxplot(df.Age)
```

&lt;Axes: &gt;



```
Q1 = df['Age'].quantile(0.25)
Q3 = df['Age'].quantile(0.75)
```

```
IQR = Q3 - Q1
```

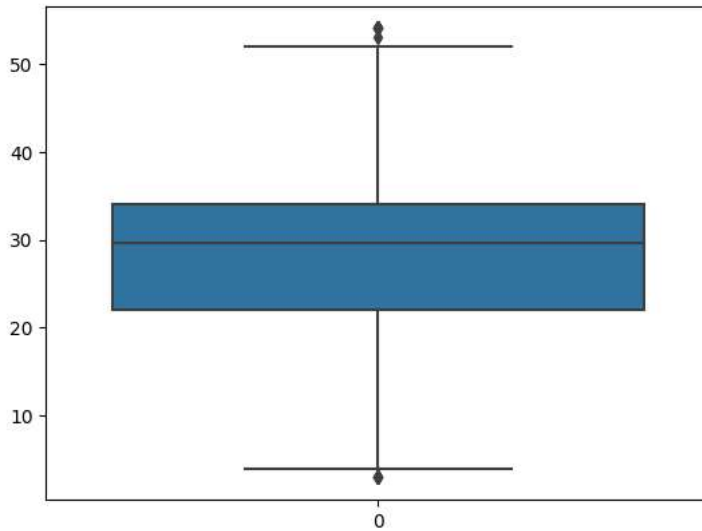
```
threshold = 1.5 * IQR
```

```
df = df[(df['Age'] >= Q1 - threshold) & (df['Age'] <= Q3 + threshold)]
```



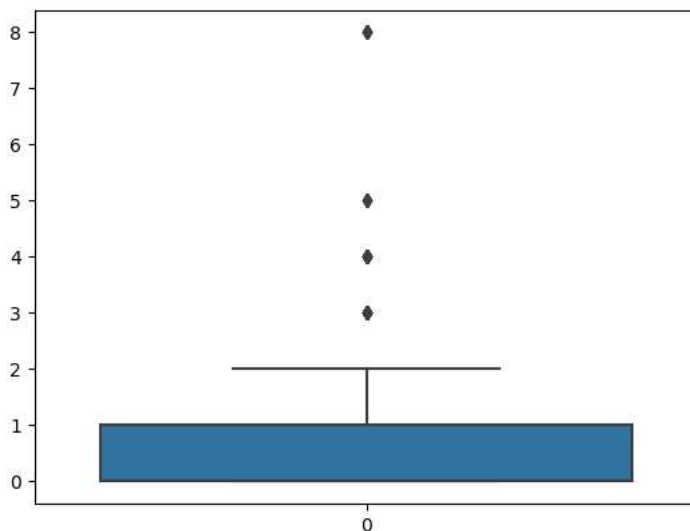
```
sns.boxplot(df.Age)
```

&lt;Axes: &gt;



```
sns.boxplot(df.SibSp)
```

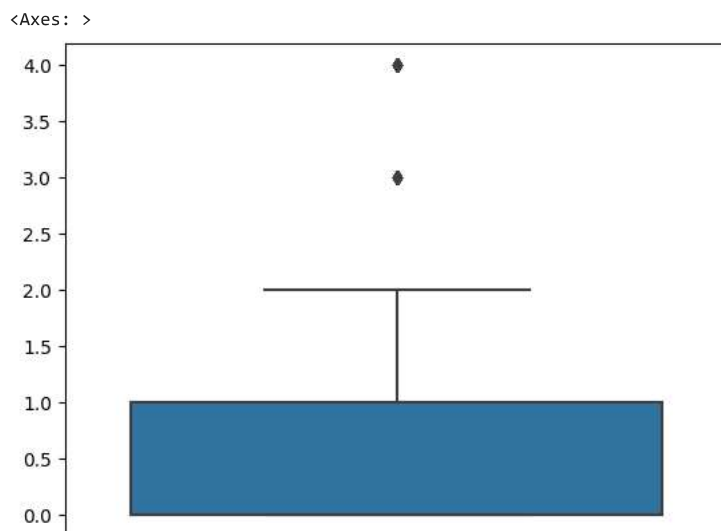
&lt;Axes: &gt;



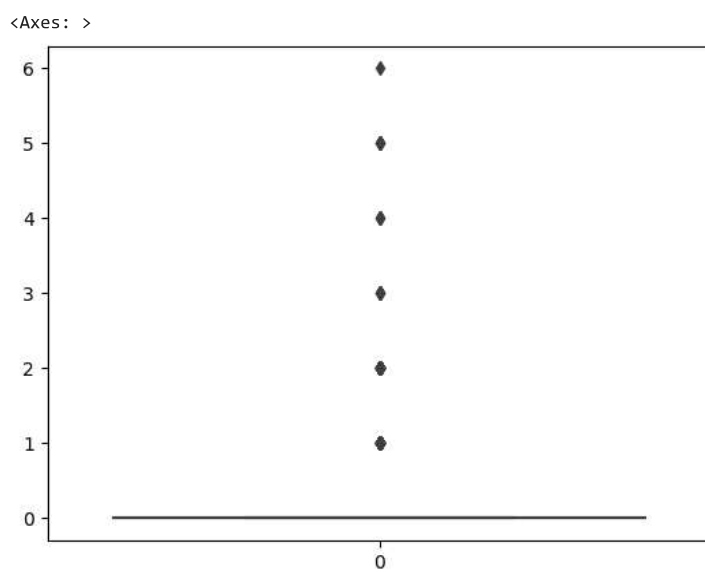
```
p99 = df.SibSp.quantile(0.99)
```

```
df = df[df.SibSp < p99]
```

```
sns.boxplot(df.SibSp)
```



```
sns.boxplot(df.Parch)
```



```
p99 = df.Parch.quantile(0.99)
```

```
df = df[df.Parch < p99]
```

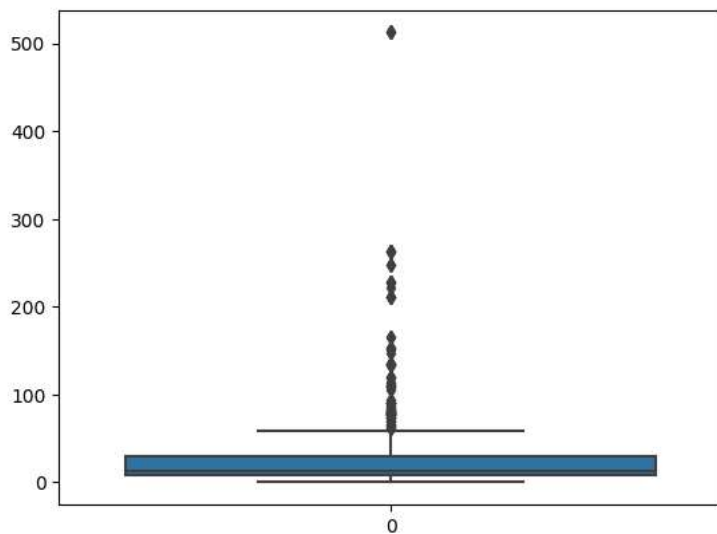
```
sns.boxplot(df["Parch"])
```



&lt;Axes: &gt;

sns.boxplot(df["Fare"])

&lt;Axes: &gt;



Q1 = df['Fare'].quantile(0.25)

Q3 = df['Fare'].quantile(0.75)

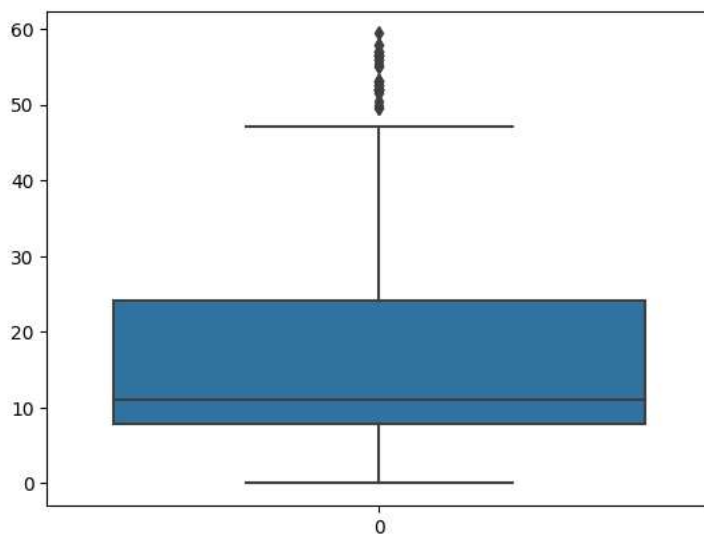
IQR = Q3 - Q1

threshold = 1.5 \* IQR

df = df[(df['Fare'] &gt;= Q1 - threshold) &amp; (df['Fare'] &lt;= Q3 + threshold)]

sns.boxplot(df.Fare)

&lt;Axes: &gt;



### Splitting Dependent and Independent Variables

x = df.drop(columns=["Survived", "PassengerId", "Name", "Ticket", "Cabin"], axis=1) # Independent variables should be in df or 2d array

x.head()

```

      Pclass    Sex      Age  SibSp  Parch    Fare  Embarked
0         3   male  22.000000     1     0   7.2500        S
2         3  female  26.000000     0     0   7.9250        S

y = pd.Series(df["Survived"])
#      0   1   0   1   0   0   0   0   0   0
y.head()

0     0
2     1
3     1
4     0
5     0
Name: Survived, dtype: int64

```

## Encoding

```

from sklearn.preprocessing import LabelEncoder

le = LabelEncoder()

x["Sex"] = le.fit_transform(x["Sex"])

x.head()

```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	1	22.000000	1	0	7.2500	S
2	3	0	26.000000	0	0	7.9250	S
3	1	0	35.000000	1	0	53.1000	S
4	3	1	35.000000	0	0	8.0500	S
5	3	1	29.699118	0	0	8.4583	Q

```

print(le.classes_)

['female' 'male']

mapping=dict(zip(le.classes_,range(len(le.classes_))))
mapping

{'female': 0, 'male': 1}

le1 = LabelEncoder()

x["Embarked"] = le1.fit_transform(x["Embarked"])

x.head()

```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	3	1	22.000000	1	0	7.2500	2
2	3	0	26.000000	0	0	7.9250	2
3	1	0	35.000000	1	0	53.1000	2
4	3	1	35.000000	0	0	8.0500	2
5	3	1	29.699118	0	0	8.4583	1

```

print(le1.classes_)

['C' 'Q' 'S']

```

```
mapping1=dict(zip(le1.classes_,range(len(le1.classes_))))
mapping1



{'C': 0, 'Q': 1, 'S': 2}
```

## Feature Scaling

```
from sklearn.preprocessing import MinMaxScaler
ms = MinMaxScaler()

x_Scaled = pd.DataFrame(ms.fit_transform(x),columns = x.columns)

x_Scaled.head()
```

	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked	
0	1.0	1.0	0.372549	0.25	0.0	0.122054	1.0	
1	1.0	0.0	0.450980	0.00	0.0	0.133418	1.0	
2	0.0	0.0	0.627451	0.25	0.0	0.893939	1.0	
3	1.0	1.0	0.627451	0.00	0.0	0.135522	1.0	
4	1.0	1.0	0.523512	0.00	0.0	0.142396	0.5	

## Splitting Training and Testing Data

```
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_state =0)

print(x_train.shape,x_test.shape,y_train.shape,y_test.shape)

(562, 7) (141, 7) (562,) (141,)
```