In [1]: import pandas as pd
 df = pd.read_csv('titanic.csv')
 df

Out[1]:		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Ci
	0	1	0	3	Braund, Mr. Owen Harris	male	22.0	100	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	100	0	113803	53.1000	C
	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	male	26.0	0	0	111369	30.0000	C
	890	891	0	3	Dooley, Mr. Patrick	male	32.0	0	0	370376	7.7500	

891 rows × 12 columns

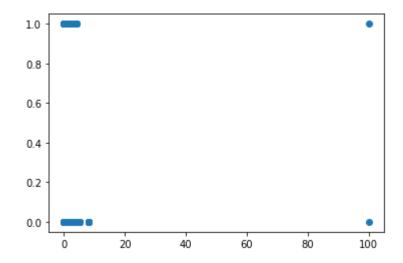
4

```
In [2]: df.isna().sum()
Out[2]: PassengerId
                         0
        Survived
                         0
        Pclass
                         0
                         0
        Name
        Sex
                         0
                       177
        Age
        SibSp
                         0
        Parch
                         0
        Ticket
                         0
                         0
        Fare
        Cabin
                       687
                         2
        Embarked
        dtype: int64
In [3]: mean_age = round(df['Age'].mean())
        mean_age
Out[3]: 30
In [4]: df['Age'].fillna(mean_age, inplace=True)
```

```
In [7]: from sklearn.preprocessing import LabelEncoder
          le = LabelEncoder()
          df['Cabin'] = le.fit_transform(df['Cabin'])
          df['Embarked'] = le.fit transform(df['Embarked'])
          df['Ticket'] = le.fit_transform(df['Ticket'])
          df['Sex'] = le.fit_transform(df['Sex'])
          df['Name'] = le.fit transform(df['Name'])
          df
 Out[7]:
                Passengerld Survived Pclass Name Sex Age SibSp Parch Ticket
                                                                                     Fare Cabin Emb
             0
                         1
                                   0
                                          3
                                               108
                                                        22.0
                                                                100
                                                                         0
                                                                              523
                                                                                   7.2500
                                                                                             147
             1
                         2
                                          1
                                                        38.0
                                   1
                                               190
                                                      0
                                                                  1
                                                                         0
                                                                              596 71.2833
                                                                                             81
             2
                         3
                                   1
                                                        26.0
                                                                  0
                                                                                             147
                                          3
                                               353
                                                      0
                                                                         0
                                                                              669
                                                                                   7.9250
             3
                         4
                                   1
                                          1
                                               272
                                                        35.0
                                                                100
                                                                         0
                                                                                 53.1000
                                                                                             55
             4
                         5
                                   0
                                          3
                                                15
                                                        35.0
                                                                  0
                                                                         0
                                                                              472
                                                                                   8.0500
                                                                                             147
                                                      1
                                  ...
                                                ...
                                                                               ...
             •••
                         ...
                                          ...
                                                          ...
                                                                        ---
                                                                                       ...
                                                                                              ...
           886
                        887
                                   0
                                          2
                                               548
                                                        27.0
                                                                         0
                                                                              101
                                                                                 13.0000
                                                                                             147
                                                      1
                                                                  0
           887
                        888
                                   1
                                          1
                                               303
                                                        19.0
                                                                  0
                                                                         0
                                                                               14 30.0000
                                                                                             30
                        889
                                   0
                                          3
                                                                         2
           888
                                               413
                                                        30.0
                                                                              675 23.4500
                                                                                             147
           889
                        890
                                   1
                                          1
                                                81
                                                        26.0
                                                                  0
                                                                         0
                                                                                  30.0000
                                                                                             60
           890
                        891
                                   0
                                          3
                                               220
                                                      1
                                                        32.0
                                                                  0
                                                                         0
                                                                              466
                                                                                   7.7500
                                                                                             147
          891 rows × 12 columns
 In [8]: mode cabin = df['Cabin'].mode()[0]
          mode_cabin
 Out[8]: 147
 In [9]: df['Cabin'].fillna(mode_cabin, inplace=True)
In [10]: mode_embarked = df['Embarked'].mode()[0]
          mode embarked
Out[10]: 2
In [12]: df['Embarked'].fillna(mode embarked, inplace=True)
```

```
In [13]: df.isna().sum()
Out[13]: PassengerId
                         0
         Survived
                         0
         Pclass
                         0
         Name
                         0
                         0
         Sex
                         0
         Age
         SibSp
                         0
         Parch
                         0
         Ticket
                         0
         Fare
                         0
         Cabin
                         0
         Embarked
                         0
         dtype: int64
In [17]: import matplotlib.pyplot as plt
         %matplotlib inline
         plt.scatter(df['SibSp'], df['Survived'])
```

Out[17]: <matplotlib.collections.PathCollection at 0x21309780e80>



C:\Users\USER\AppData\Local\Temp\ipykernel_10164\4235250280.py:4: SettingWithCo
pyWarning:

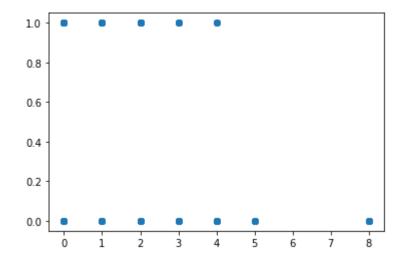
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
df['SibSp'][x] = mode_sibsp
```

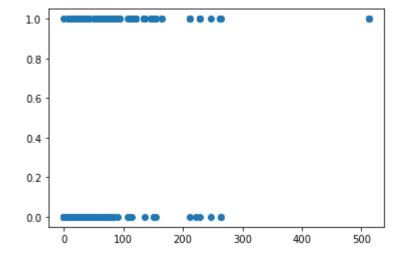
```
In [20]: import matplotlib.pyplot as plt
%matplotlib inline
plt.scatter(df['SibSp'], df['Survived'])
```

Out[20]: <matplotlib.collections.PathCollection at 0x21309acd490>



```
In [22]: import matplotlib.pyplot as plt
%matplotlib inline
    plt.scatter(df['Fare'], df['Survived'])
```

Out[22]: <matplotlib.collections.PathCollection at 0x21309b8ae20>



```
In [23]: mode_fare = df['Fare'].mode()[0]
for x in range(0,891):
    if df['Fare'][x] >= 400:
        df['Fare'][x] = mode_fare
```

C:\Users\USER\AppData\Local\Temp\ipykernel_10164\690323287.py:4: SettingWithCop
yWarning:

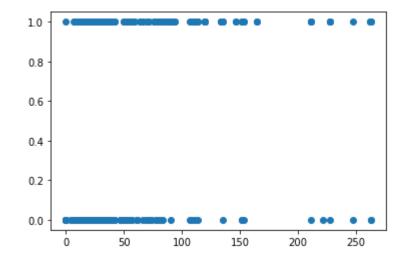
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

df['Fare'][x] = mode_fare

```
In [24]: import matplotlib.pyplot as plt
%matplotlib inline
   plt.scatter(df['Fare'], df['Survived'])
```

Out[24]: <matplotlib.collections.PathCollection at 0x21309c0a220>



```
In [25]: x = df.drop(['Survived'], axis=1)
y = df['Survived']
```

- In [26]: from sklearn.model_selection import train_test_split
 x_train, x_test, y_train, y_test = train_test_split(x,y,test_size=0.2,random_stat
- In [27]: from sklearn.preprocessing import MinMaxScaler
 min_max_scaler = MinMaxScaler()

```
In [28]: x_train = min_max_scaler.fit_transform(x_train)
x_test = min_max_scaler.fit_transform(x_test)
```

```
In [29]: from sklearn.linear model import LogisticRegression
         model = LogisticRegression()
In [30]: model.fit(x_train,y_train)
Out[30]: LogisticRegression()
In [31]: y_pred = model.predict(x_test)
         y pred
Out[31]: array([0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0,
                1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0,
                1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1,
                0, 0, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1,
                0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1,
                1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0,
                0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0,
                0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0,
                0, 1, 0], dtype=int64)
In [33]: from sklearn.metrics import classification_report
         classification_report(y_test, y_pred)
Out[33]: '
                       precision
                                    recall f1-score
                                                       support\n\n
                                                                                    0.
                                                                                0.76
         81
                 0.90
                          0.85
                                     105\n
                                                     1
                                                             0.83
                                                                       0.70
                                                     0.82
                                                                179\n
         74\n\n
                   accuracy
                                                                       macro avg
```

179\nweighted avg

0.82

0.82

0.81

0.82

179\n'

0.80

0.80