

## Data Wrangle Report

This project is about the Titanic dataset which contains information about the passengers who were on it while it sank. For more details, look at

[https://en.wikipedia.org/wiki/Passengers\\_of\\_the\\_RMS\\_Titanic](https://en.wikipedia.org/wiki/Passengers_of_the_RMS_Titanic)

(you can gather it from

<https://www.datacamp.com/community/tutorials/k-means-clustering-python>).

The original data set contains 1039 rows with 21 attributes (columns)

```
: data = pd.read_csv("dataBeforeCleaning.csv")
: data.head(2)
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	...	Embarked	WikiId	Name_wiki	Age_wiki	Hometown	Boarded
0	1	0.0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	...	S	691.0	Braund, Mr. Owen Harris	22.0	Bridgerule, Devon, England	Southampton
1	2	1.0	1	Cumings, Mrs. John Bradley (Florence Briggs Thayer)	female	38.0	1	0	PC 17599	71.2833	...	C	90.0	Cumings, Mrs. Florence Briggs (née Thayer)	35.0	New York, New York, US	Cherbourg

2 rows x 21 columns

```
: data.shape
: (1309, 21)
```

Now, we will discuss about Cleaning :

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

1<sup>st</sup> | we read our data.

```
data = pd.read_csv("dataBeforeCleaning.csv")
```

## Assessment & issues:

**tidy issue:** the column name contain two values first value is surname and second value is full name

To solve this issue we must split name column into 2 columns

1<sup>st</sup> column is surname , 2<sup>nd</sup> column is full name

But first before starting data preprocessing operations we must take a copy of this data

```
data_copy = data.copy()
```

Now, we split the data column then remove it.

```
In [47]: data_copy[["surname", "full name"]] = data_copy["Name"].str.split(",", 1, expand = True)

In [48]: data_copy.drop("Name", axis = 1, inplace = True)
```

output after splitting this column into 2 columns:

```
In [49]: list(data_copy)

Out[49]: ['PassengerId',
'Survived',
'Pclass',
'Sex',
'Age',
'SibSp',
'Parch',
'Ticket',
'Fare',
'Cabin',
'Embarked',
'WikiId',
'Name_wiki',
'Age_wiki',
'Hometown',
'Boarded',
'Destination',
'Lifeboat',
'Body',
'Class',
'surname',
'full name']
```

**Quality issues:**

**Second issue:** there are many duplicates and redundancy columns like:

Age and Age\_Wiki

Name and Name Wiki

Passengers Id and Wiki Id

**Solution:**

we should drop the duplicates columns:

```
data_copy.drop(["Name_wiki", "Age_wiki", "WikiId", "Class"], axis = 1, inplace = True)
```

the output of new lists after dropping the duplicates columns:

```
list(data_copy)
['PassengerId',
 'Survived',
 'Pclass',
 'Sex',
 'Age',
 'SibSp',
 'Parch',
 'Ticket',
 'Fare',
 'Cabin',
 'Embarked',
 'Hometown',
 'Boarded',
 'Destination',
 'Lifeboat',
 'Body',
 'surname',
 'full name']
```

### **Third Issue:**

There are missing values in Age column

```
: data_copy["Age"].isnull().sum()
: 263
```

### **Solution:**

Since the age maybe an effective parameter we should fill the missing value with the column's mean

```
In [53]: data_copy["Age"].fillna(data_copy["Age"].mean,inplace = True)
```

then after doing that we find the missing values in this column is 0

```
In [54]: data_copy["Age"].isnull().sum()
Out[54]: 0
```

### **Fourth Issue:**

most of cabin column is missing data

```
data_copy.head(1).sum()

PassengerId      0
Survived         418
Pclass           0
Sex              0
Age             0
SibSp           0
Parch           0
Ticket           0
Fare             1
Cabin          1014
```

### Solution:

Since it is not important feature we can drop cabin column , then the new list will be:

```
In [14]: data_copy.drop("Cabin",axis = 1,inplace = True)
```

```
In [58]: list(data_copy)
```

```
Out[58]: ['PassengerId',
          'Survived',
          'Pclass',
          'Sex',
          'Age',
          'SibSp',
          'Parch',
          'Ticket',
          'Fare',
          'Embarked',
          'Hometown',
          'Boarded',
          'Destination',
          'Lifeboat',
          'Body',
          'Class',
          'surname',
          'full name']
```

the fourth issue

### Fifth Issue:

There are missing values in lifeboat & body columns:

### Solution:

Drop lifeboat and body columns:

### Sixth Issue:

Sibsp and Parch column Names are not interpretable

### Solution:

Rename SibSp column with Number of Siblings and parch to number of parents and children

```
data_copy.rename(columns={"SibSp": "Number of siblings", "Parch": "Number of parents and children"}, inplace=True)
data_copy.head()
```

PassengerId	Survived	Pclass	Sex	Age	Number of siblings	Number of parents and children	Ticket	Fare	Cabin	Embarked	Hometown	Boarded	Destination	Lifeboat
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Seventh issue:

data has columns which contain string data and k mean clustering can only handling numeric data

Column	non-null count	dtype
PassengerId	1309 non-null	int64
Survived	891 non-null	float64
Pclass	1309 non-null	int64
Sex	1309 non-null	object
Age	1309 non-null	float64
Number of siblings	1309 non-null	int64
Number of parents and children	1309 non-null	int64
Ticket	1309 non-null	object
Fare	1308 non-null	float64
Cabin	295 non-null	object
0 Embarked	1307 non-null	object
1 Hometown	1304 non-null	object
2 Boarded	1304 non-null	object
3 Destination	1304 non-null	object
4 Lifeboat	502 non-null	object
5 Body	130 non-null	object
6 surname	1309 non-null	object
7 full name	1309 non-null	object

types: float64(3), int64(4), object(11)  
memory usage: 184.2+ KB

Solution:

We will convert columns that what we will use in the model into numeric using Label Encoder class from sklearn library we will need only "Sex" column so we will encode this column and drop other string columns.

```
from sklearn.preprocessing import LabelEncoder
LE = LabelEncoder()
LE.fit(data_copy["Sex"])
```

```
LabelEncoder()
```

```
data_copy["Sex"] = LE.transform(data_copy["Sex"])
```

### Seventh Issue:

Survived column has many missing values

```
2]: data_copy.isnull().sum()
```

```
2]: Survived      418
    Age           0
    Pclass        0
    Sex           0
```

### Solution:

Since we have done most of the data cleaning work and we choiced the columns that we will need we will separate our data into two separated files first file will contain our cleaned data without rows which contain null values and we will use this data to train our clustering model to predict the missing values in Survived column in the second file(test\_data)

```
In [20]: data_train = data_copy[data_copy.Survived.isnull() == False]
        data_test = data_copy[data_copy.Survived.isnull()]
        data_test.shape
```

```
Out[20]: (418, 16)
```

