

Group A

Assignment 1

Data Wrangling I

Import all the required Python Libraries.

```
# code here
import numpy as np
import pandas as pd
```

Load the Dataset into pandas dataframe.

```
# code here
df = pd.read_csv("Titanic.csv")
```

```
# code here
df.head()
```

	sex	age	sibsp	parch	fare	embarked	class	who	alone
survived									
0	male	22.0	1	0	7.2500	S	Third	man	False
0									
1	female	38.0	1	0	71.2833	C	First	woman	False
1									
2	female	26.0	0	0	7.9250	S	Third	woman	True
1									
3	female	35.0	1	0	53.1000	S	First	woman	False
1									
4	male	35.0	0	0	8.0500	S	Third	man	True
0									

```
df.sample()
```

	sex	age	sibsp	parch	fare	embarked	class	who	alone
survived									
654	female	18.0	0	0	6.75	Q	Third	woman	True
0									

```
df.tail()
```

	sex	age	sibsp	parch	fare	embarked	class	who	alone
\									
886	male	27.0	0	0	13.00	S	Second	man	True

887	female	19.0	0	0	30.00	S	First	woman	True
888	female	NaN	1	2	23.45	S	Third	woman	False
889	male	26.0	0	0	30.00	C	First	man	True
890	male	32.0	0	0	7.75	Q	Third	man	True

	survived
886	0
887	1
888	0
889	1
890	0

Data Preprocessing

check for missing values in the data using pandas isnull()

```
# to highlight esc+m
# we use # as h1,## as h2 and so on
```

```
df.isnull()
```

	sex	age	sibsp	parch	fare	embarked	class	who	alone
\									
0	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False
..
886	False	False	False	False	False	False	False	False	False
887	False	False	False	False	False	False	False	False	False
888	False	True	False	False	False	False	False	False	False
889	False	False	False	False	False	False	False	False	False
890	False	False	False	False	False	False	False	False	False

	survived
--	----------

```
0      False
1      False
2      False
3      False
4      False
..      ...
886     False
887     False
888     False
889     False
890     False
```

```
[891 rows x 10 columns]
```

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

```
sex      0
age     177
sibsp    0
parch    0
fare     0
embarked  2
class    0
who      0
alone    0
survived  0
dtype: int64
```

```
df["age"].fillna(df["age"].mean(),inplace=True)
# if changes are seen after execution then these changes are
temporary to do it permanent we use inplace
# to check these check above run isnull function
```

```
df["embarked"].value_counts()
```

```
embarked
S      644
C      168
Q       77
Name: count, dtype: int64
```

```
df["embarked"].fillna('S')
```

```
0      S
1      C
2      S
3      S
4      S
..      ..
886     S
887     S
```

```
888    S
889    C
890    Q
Name: embarked, Length: 891, dtype: object
```

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

```
sex      0
age      0
sibsp    0
parch    0
fare     0
embarked  2
class    0
who      0
alone    0
survived  0
dtype: int64
```

```
df["embarked"].fillna('S',inplace=True)
```

```
df.isna().sum()
```

```
sex      0
age      0
sibsp    0
parch    0
fare     0
embarked  0
class    0
who      0
alone    0
survived  0
dtype: int64
```

describe() function to get some initial statistics. Provide variable descriptions.

```
# code here
df.describe()
```

	age	sibsp	parch	fare	survived
count	891.000000	891.000000	891.000000	891.000000	891.000000
mean	29.699118	0.523008	0.381594	32.204208	0.383838
std	13.002015	1.102743	0.806057	49.693429	0.486592
min	0.420000	0.000000	0.000000	0.000000	0.000000
25%	22.000000	0.000000	0.000000	7.910400	0.000000
50%	29.699118	0.000000	0.000000	14.454200	0.000000
75%	35.000000	1.000000	0.000000	31.000000	1.000000
max	80.000000	8.000000	6.000000	512.329200	1.000000

```
df.mean()
```

```

sex          0.647587
age          29.699118
sibsp        0.523008
parch        0.381594
fare         32.204208
embarked     1.536476
class        1.308642
who          1.210999
alone        0.602694
survived     0.383838
dtype: float64

df["age"].quantile(0.25)

22.0

```

Types of variables

```

# code here
df.dtypes

sex          object
age          float64
sibsp        int64
parch        int64
fare         float64
embarked     object
class        object
who          object
alone        bool
survived     int64
dtype: object

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 10 columns):
 #   Column      Non-Null Count  Dtype
---  -
 0   sex         891 non-null   object
 1   age         891 non-null   float64
 2   sibsp       891 non-null   int64
 3   parch       891 non-null   int64
 4   fare        891 non-null   float64
 5   embarked    891 non-null   object
 6   class       891 non-null   object
 7   who         891 non-null   object
 8   alone       891 non-null   bool
 9   survived    891 non-null   int64

```

```
dtypes: bool(1), float64(2), int64(3), object(4)
memory usage: 63.6+ KB
```

```
df["age"].sample(10)
```

```
90      29.000000
126     29.699118
857     51.000000
164      1.000000
842     30.000000
475     29.699118
889     26.000000
668     43.000000
319     40.000000
229     29.699118
Name: age, dtype: float64
```

Check the dimensions of the data frame

```
# code here
```

```
df.shape
```

```
(891, 10)
```

Data Formatting and Data Normalization

Summarize the types of variables by checking the data types (i.e., character, numeric, integer, factor, and logical) of the variables in the data set.

```
# code here
```

```
df.nunique()
```

```
sex      2
age     89
sibsp    7
parch    7
fare    248
embarked  3
class    3
who       3
alone    2
survived  2
dtype: int64
```

```
df["sex"].value_counts()
```

```
sex
male    577
female  314
Name: count, dtype: int64
```

```
df["embarked"].value_counts()
```

```
embarked
```

```
S      646
```

```
C      168
```

```
Q       77
```

```
Name: count, dtype: int64
```

```
df["sibsp"].value_counts()
```

```
sibsp
```

```
0      608
```

```
1      209
```

```
2       28
```

```
4       18
```

```
3       16
```

```
8        7
```

```
5        5
```

```
Name: count, dtype: int64
```

```
df["parch"].value_counts()
```

```
parch
```

```
0      678
```

```
1      118
```

```
2       80
```

```
5        5
```

```
3        5
```

```
4        4
```

```
6        1
```

```
Name: count, dtype: int64
```

```
df["class"].value_counts()
```

```
class
```

```
Third     491
```

```
First     216
```

```
Second    184
```

```
Name: count, dtype: int64
```

```
df["who"].value_counts()
```

```
who
```

```
man      537
```

```
woman    271
```

```
child     83
```

```
Name: count, dtype: int64
```

```
df["alone"].value_counts()
```

```
alone
```

```
True     537
```

```
False    354
Name: count, dtype: int64

df["survived"].value_counts()

survived
0     549
1     342
Name: count, dtype: int64
```

If variables are not in the correct data type, apply proper type conversions.

```
# code here
# df.age.astype('int64')
```

Turn categorical variables into quantitative variables in Python.

```
# code here
# replace function takes two parameter i.) list of string
# ii.) list of numbers to replace

df["sex"].replace(['female', 'male'], [0, 1], inplace=True)
df["who"].replace(['child', 'man', 'woman'], [0, 1, 2], inplace=True)
df["embarked"].replace(['C', 'Q', 'S'], [0, 1, 2], inplace=True)
df["class"].replace(['First', 'Second', 'Third'], [0, 1, 2], inplace=True)
df["alone"].replace(['False', 'True'], [0, 1], inplace=True)
```

```
df.dtypes
```

```
sex          int64
age          float64
sibsp        int64
parch        int64
fare         float64
embarked     int64
class        int64
who          int64
alone        bool
survived     int64
dtype: object
```

```
df.describe()
```

	sex	age	sibsp	parch	fare
embarked \					
count	891.000000	891.000000	891.000000	891.000000	891.000000
mean	0.647587	29.699118	0.523008	0.381594	32.204208


```

1.536476
std      0.477990    13.002015    1.102743    0.806057    49.693429
0.791503
min      0.000000     0.420000    0.000000    0.000000    0.000000
0.000000
25%      0.000000    22.000000    0.000000    0.000000    7.910400
1.000000
50%      1.000000    29.699118    0.000000    0.000000    14.454200
2.000000
75%      1.000000    35.000000    1.000000    0.000000    31.000000
2.000000
max      1.000000    80.000000    8.000000    6.000000    512.329200
2.000000

```

```

              class      who      survived
count  891.000000  891.000000  891.000000
mean    1.308642    1.210999    0.383838
std     0.836071    0.594291    0.486592
min     0.000000    0.000000    0.000000
25%     1.000000    1.000000    0.000000
50%     2.000000    1.000000    0.000000
75%     2.000000    2.000000    1.000000
max     2.000000    2.000000    1.000000

```

```
df["age"].unique()
```

```

array([22.          , 38.          , 26.          , 35.          ,
29.69911765,
54.          , 2.          , 27.          , 14.          ,
4.          ,
58.          , 20.          , 39.          , 55.          ,
31.          ,
34.          , 15.          , 28.          , 8.          ,
19.          ,
40.          , 66.          , 42.          , 21.          ,
18.          ,
3.          , 7.          , 49.          , 29.          ,
65.          ,
28.5         , 5.          , 11.          , 45.          ,
17.          ,
32.          , 16.          , 25.          , 0.83         ,
30.          ,
33.          , 23.          , 24.          , 46.          ,
59.          ,
71.          , 37.          , 47.          , 14.5         ,
70.5         ,
32.5         , 12.          , 9.          , 36.5         ,
51.          ,
55.5         , 40.5         , 44.          , 1.          ,
61.          ,

```

```

20.5 56. , 50. , 36. , 45.5 ,
23.5 62. , 41. , 52. , 63. ,
64. 0.92 , 43. , 60. , 10. ,
57. 13. , 48. , 0.75 , 53. ,
0.67 80. , 70. , 24.5 , 6. ,
30.5 , 0.42 , 34.5 , 74. ])
```

```
df.sample(10)
```

	sex	age	sibsp	parch	fare	embarked	class	who
alone \								
293	0	24.000000	0	0	8.8500	2	2	2
True								
729	0	25.000000	1	0	7.9250	2	2	2
False								
685	1	25.000000	1	2	41.5792	0	1	1
False								
677	0	18.000000	0	0	9.8417	2	2	2
True								
288	1	42.000000	0	0	13.0000	2	1	1
True								
571	0	53.000000	2	0	51.4792	2	0	2
False								
356	0	22.000000	0	1	55.0000	2	0	2
False								
815	1	29.699118	0	0	0.0000	2	0	1
True								
334	0	29.699118	1	0	133.6500	2	0	2
False								
238	1	19.000000	0	0	10.5000	2	1	1
True								

	survived
293	0
729	0
685	0
677	1
288	1
571	1
356	1
815	0
334	1
238	0