

Practical No.1

Data Science and Visualization (Honors Course)

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Class: TE ENT C 'B'

In this practical we will access an open source dataset 'titanic.csv' and apply pre-processing techniques on the raw dataset.

In [1]:

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

We will now check the current version of all the packages which we imported.

In [2]:

```
pd.__version__
```

Out[2]:

```
'1.2.4'
```

In [3]:

```
np.__version__
```

Out[3]:

```
'1.20.1'
```

In [4]:

```
sns.__version__
```

Out[4]:

```
'0.11.1'
```

We will now get the datasets which are already inbuilt in the packages.

In [5]:

```
sns.get_dataset_names()
```

Out[5]:

```
['anagrams',  
'anscombe',  
'attention',  
'brain_networks',  
'car_crashes',  
'diamonds',  
'dots',  
'exercise',  
'flights',  
'fmri',  
'gammas',  
'geyser',  
'iris',  
'mpg',  
'penguins',  
'planets',  
'taxis',  
'tips',  
'titanic']
```

There are various ways to import a dataset which are as follows:

In [6]:

```
dataset = sns.load_dataset('titanic')
```

In [7]:

```
dataset
```

Out[7]:

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_n
0	0	3	male	22.0	1	0	7.2500	S	Third	man	1
1	1	1	female	38.0	1	0	71.2833	C	First	woman	Fi
2	1	3	female	26.0	0	0	7.9250	S	Third	woman	Fi
3	1	1	female	35.0	1	0	53.1000	S	First	woman	Fi
4	0	3	male	35.0	0	0	8.0500	S	Third	man	1
...	
886	0	2	male	27.0	0	0	13.0000	S	Second	man	1
887	1	1	female	19.0	0	0	30.0000	S	First	woman	Fi
888	0	3	female	NaN	1	2	23.4500	S	Third	woman	Fi
889	1	1	male	26.0	0	0	30.0000	C	First	man	1
890	0	3	male	32.0	0	0	7.7500	Q	Third	man	1

891 rows × 15 columns



In [8]:

```
df = pd.read_csv('https://web.stanford.edu/class/archive/cs/cs109/cs109.1166/stuff/titanic.
```

In [9]:

df

Out[9]:

	Survived	Pclass	Name	Sex	Age	Siblings/Spouses Aboard	Parents/Children Aboard	Fare
0	0	3	Mr. Owen Harris Braund	male	22.0	1	0	7.2500
1	1	1	Mrs. John Bradley (Florence Briggs Thayer) Cum...	female	38.0	1	0	71.2833
2	1	3	Miss. Laina Heikkinen	female	26.0	0	0	7.9250
3	1	1	Mrs. Jacques Heath (Lily May Peel) Futrelle	female	35.0	1	0	53.1000
4	0	3	Mr. William Henry Allen	male	35.0	0	0	8.0500
...
882	0	2	Rev. Juozas Montvila	male	27.0	0	0	13.0000
883	1	1	Miss. Margaret Edith Graham	female	19.0	0	0	30.0000
884	0	3	Miss. Catherine Helen Johnston	female	7.0	1	2	23.4500
885	1	1	Mr. Karl Howell Behr	male	26.0	0	0	30.0000
886	0	3	Mr. Patrick Dooley	male	32.0	0	0	7.7500

887 rows × 8 columns

We will now perform certain pre processing operations on our dataset.

In [11]:

```
df.columns #The title of all the columns in the dataset.
```

Out[11]:

```
Index(['Survived', 'Pclass', 'Name', 'Sex', 'Age', 'Siblings/Spouses Aboard',  
      'Parents/Children Aboard', 'Fare'],  
      dtype='object')
```

In [12]:

```
df.shape
```

Out[12]:

```
(887, 8)
```

In [13]:

```
dataset.shape
```

Out[13]:

```
(891, 15)
```

In [14]:

```
dataset.columns
```

Out[14]:

```
Index(['survived', 'pclass', 'sex', 'age', 'sibsp', 'parch', 'fare',  
      'embarked', 'class', 'who', 'adult_male', 'deck', 'embark_town',  
      'alive', 'alone'],  
      dtype='object')
```

In [16]:

```
df.head() #the .head() function returns the first five rows of dataset by default.
```

Out[16]:

	Survived	Pclass	Name	Sex	Age	Siblings/Spouses Aboard	Parents/Children Aboard	Fare
0	0	3	Mr. Owen Harris Braund	male	22.0	1	0	7.2500
1	1	1	Mrs. John Bradley (Florence Briggs Thayer) Cum...	female	38.0	1	0	71.2833
2	1	3	Miss. Laina Heikkinen	female	26.0	0	0	7.9250
3	1	1	Mrs. Jacques Heath (Lily May Peel) Futrelle	female	35.0	1	0	53.1000
4	0	3	Mr. William Henry Allen	male	35.0	0	0	8.0500

In [17]:

```
dataset.head()
```

Out[17]:

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

In [19]:

```
df.tail() #the .tail() function returns the last five rows of dataset by default.
```

Out[19]:

	Survived	Pclass	Name	Sex	Age	Siblings/Spouses Aboard	Parents/Children Aboard	Fare
882	0	2	Rev. Juozas Montvila	male	27.0	0	0	13.00
883	1	1	Miss. Margaret Edith Graham	female	19.0	0	0	30.00
884	0	3	Miss. Catherine Helen Johnston	female	7.0	1	2	23.45
885	1	1	Mr. Karl Howell Behr	male	26.0	0	0	30.00
886	0	3	Mr. Patrick Dooley	male	32.0	0	0	7.75

In [20]:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 887 entries, 0 to 886
Data columns (total 8 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Survived                             887 non-null    int64
1   Pclass                               887 non-null    int64
2   Name                                 887 non-null    object
3   Sex                                  887 non-null    object
4   Age                                  887 non-null    float64
5   Siblings/Spouses Aboard              887 non-null    int64
6   Parents/Children Aboard              887 non-null    int64
7   Fare                                  887 non-null    float64
dtypes: float64(2), int64(4), object(2)
memory usage: 55.6+ KB
```

In [21]:

dataset.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 15 columns):
#   Column          Non-Null Count  Dtype
---  -
0   survived        891 non-null    int64
1   pclass          891 non-null    int64
2   sex             891 non-null    object
3   age            714 non-null    float64
4   sibsp          891 non-null    int64
5   parch          891 non-null    int64
6   fare           891 non-null    float64
7   embarked       889 non-null    object
8   class          891 non-null    category
9   who            891 non-null    object
10  adult_male     891 non-null    bool
11  deck          203 non-null    category
12  embark_town    889 non-null    object
13  alive          891 non-null    object
14  alone          891 non-null    bool
dtypes: bool(2), category(2), float64(2), int64(4), object(5)
memory usage: 80.7+ KB
```

In [22]:

dataset.describe()

Out[22]:

	survived	pclass	age	sibsp	parch	fare
count	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000
mean	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208
std	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429
min	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000
25%	0.000000	2.000000	20.125000	0.000000	0.000000	7.910400
50%	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200
75%	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000
max	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200

In [23]:

```
df.describe()
```

Out[23]:

	Survived	Pclass	Age	Siblings/Spouses Aboard	Parents/Children Aboard	Fare
count	887.000000	887.000000	887.000000	887.000000	887.000000	887.000000
mean	0.385569	2.305524	29.471443	0.525366	0.383315	32.30542
std	0.487004	0.836662	14.121908	1.104669	0.807466	49.78204
min	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000
25%	0.000000	2.000000	20.250000	0.000000	0.000000	7.92500
50%	0.000000	3.000000	28.000000	0.000000	0.000000	14.45420
75%	1.000000	3.000000	38.000000	1.000000	0.000000	31.13750
max	1.000000	3.000000	80.000000	8.000000	6.000000	512.32920

In [24]:

```
df.count()
```

Out[24]:

```
Survived      887
Pclass        887
Name          887
Sex           887
Age           887
Siblings/Spouses Aboard  887
Parents/Children Aboard  887
Fare          887
dtype: int64
```

In [25]:

```
dataset.count()
```

Out[25]:

```
survived      891
pclass        891
sex           891
age           714
sibsp         891
parch         891
fare          891
embarked      889
class         891
who           891
adult_male    891
deck          203
embark_town   889
alive         891
alone         891
dtype: int64
```

In [26]:

```
dataset.isnull().sum()
```

Out[26]:

```
survived      0
pclass        0
sex           0
age           177
sibsp         0
parch         0
fare          0
embarked      2
class         0
who           0
adult_male    0
deck          688
embark_town    2
alive         0
alone         0
dtype: int64
```

In [27]:

```
dataset = dataset.drop('deck', axis = 1)
```

In [28]:

```
dataset.isnull().sum()
```

Out[28]:

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

In [29]:

```
dataset['age'] = dataset['age'].fillna(dataset['age'].median())
```

In [30]:

```
dataset.isnull().sum()
```

Out[30]:

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

In [31]:

```
dataset['embarked'].mode()[0]
```

Out[31]:

```
'S'
```

In [32]:

```
dataset['embark_town'].mode()[0]
```

Out[32]:

'Southampton'

In [33]:

```
dataset['embarked'] = dataset['embarked'].fillna(  
    dataset['embarked'].mode()[0])
```

In [34]:

```
dataset['embark_town'] = dataset['embark_town'].fillna(  
    dataset['embark_town'].mode()[0])
```

In [35]:

```
dataset.isnull().sum()
```

Out[35]:

```
survived      0  
pclass        0  
sex           0  
age           0  
sibsp         0  
parch         0  
fare          0  
embarked      0  
class         0  
who           0  
adult_male    0  
embark_town   0  
alive         0  
alone         0  
dtype: int64
```

In [36]:

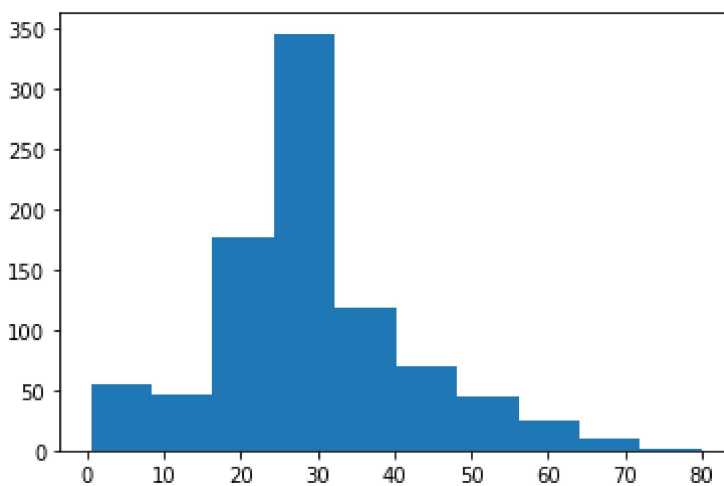
```
dataset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 14 columns):
 #   Column          Non-Null Count  Dtype  
---  -
 0   survived        891 non-null    int64  
 1   pclass          891 non-null    int64  
 2   sex             891 non-null    object  
 3   age             891 non-null    float64 
 4   sibsp          891 non-null    int64  
 5   parch          891 non-null    int64  
 6   fare            891 non-null    float64 
 7   embarked        891 non-null    object  
 8   class           891 non-null    category
 9   who             891 non-null    object  
10  adult_male      891 non-null    bool    
11  embark_town     891 non-null    object  
12  alive           891 non-null    object  
13  alone           891 non-null    bool    
dtypes: bool(2), category(1), float64(2), int64(4), object(5)
memory usage: 79.4+ KB
```

Visualization of dataset

In [37]:

```
plt.hist(dataset['age']);
```

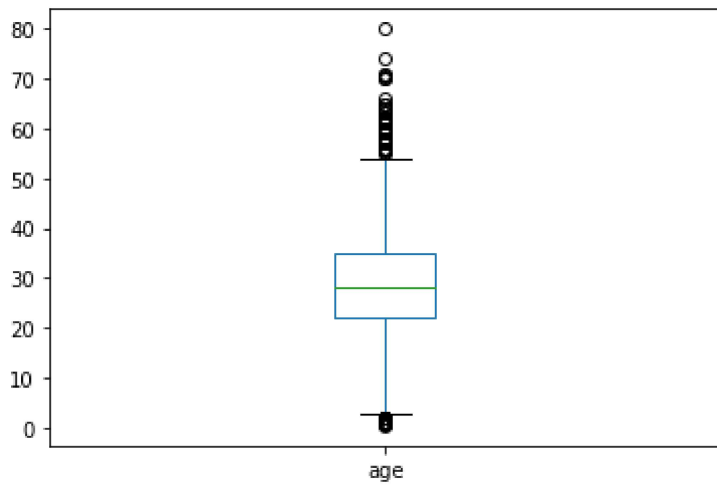


In [38]:

```
dataset['age'].plot(kind='box')
```

Out[38]:

<AxesSubplot:>

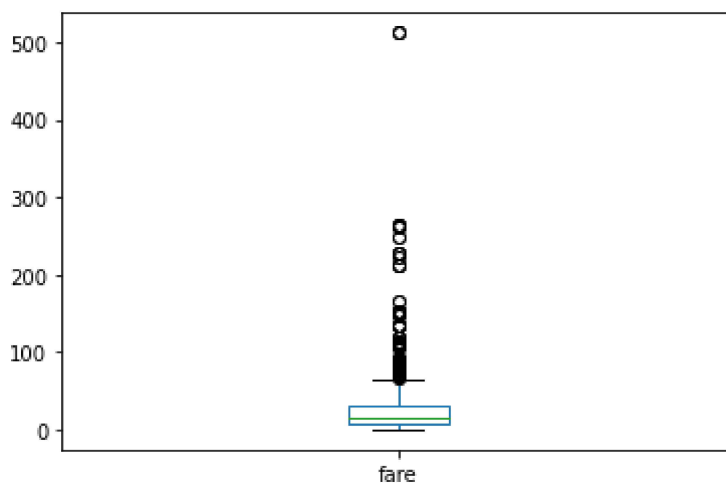


In [39]:

```
dataset['fare'].plot(kind='box')
```

Out[39]:

<AxesSubplot:>



In [40]:

dataset.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 14 columns):
#   Column          Non-Null Count  Dtype
---  -
0   survived        891 non-null    int64
1   pclass          891 non-null    int64
2   sex              891 non-null    object
3   age              891 non-null    float64
4   sibsp           891 non-null    int64
5   parch           891 non-null    int64
6   fare            891 non-null    float64
7   embarked        891 non-null    object
8   class           891 non-null    category
9   who             891 non-null    object
10  adult_male      891 non-null    bool
11  embark_town     891 non-null    object
12  alive           891 non-null    object
13  alone           891 non-null    bool
dtypes: bool(2), category(1), float64(2), int64(4), object(5)
memory usage: 79.4+ KB
```

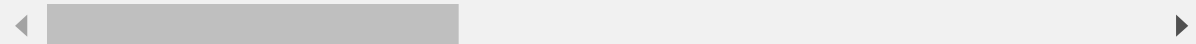
In [41]:

pd.get_dummies(dataset).head()

Out[41]:

	survived	pclass	age	sibsp	parch	fare	adult_male	alone	sex_female	sex_male	...
0	0	3	22.0	1	0	7.2500	True	False	0	1	...
1	1	1	38.0	1	0	71.2833	False	False	1	0	...
2	1	3	26.0	0	0	7.9250	False	True	1	0	...
3	1	1	35.0	1	0	53.1000	False	False	1	0	...
4	0	3	35.0	0	0	8.0500	True	True	0	1	...

5 rows × 24 columns



Training our Dataset

Importing the required library.

In [42]:

```
from sklearn.model_selection import train_test_split
```

In [43]:

```
train, test = train_test_split(dataset, test_size=0.20)
```

In [44]:

```
len(dataset)
```

Out[44]:

891

In [45]:

```
len(train)
```

Out[45]:

712

In [46]:

```
len(test)
```

Out[46]:

179

In []:

In []:

In []: