

```
In [ ]: import pandas as pd
import numpy as np
pd.set_option("display.precision", 2)

import matplotlib.pyplot as plt
import seaborn as sns
sns.set()

%config InlineBackend.figure_format = 'svg'
```

Lets import the dataset we have created first

```
In [ ]: df = pd.read_csv("./Data/data.csv", sep=",")
df.drop(['Unnamed: 0'], axis=1, inplace=True) # There were some formatting issues
                                              # writing the csv
```

## BASIC EDA

```
In [ ]: df.head()
```

```
Out[ ]:   DISTRICT  UPAZILA  STATION_ID  STATION_NAME      DATE  RAIN_FALL(mm)  LATITUDE  LONGITUDE
0  Bandarban     Lama       CL317        Lama  01-jan-2017         0.0      21.81      92.1
1  Bandarban     Lama       CL317        Lama  02-jan-2017         0.0      21.81      92.1
2  Bandarban     Lama       CL317        Lama  03-jan-2017         0.0      21.81      92.1
3  Bandarban     Lama       CL317        Lama  04-jan-2017         0.0      21.81      92.1
4  Bandarban     Lama       CL317        Lama  05-jan-2017         0.0      21.81      92.1
```

The shape of our dataset:

- we have 1826 samples containing 9 features(includes target)

```
In [ ]: df.shape
```

```
Out[ ]: (1826, 9)
```

The columns are:

```
In [ ]: df.columns
```

```
Out[ ]: Index(['DISTRICT', 'UPAZILA', 'STATION_ID', 'STATION_NAME', 'DATE',
   'RAIN_FALL(mm)', 'LATITUDE', 'LONGITUDE', 'WATER_LEVEL(m)'],
   dtype='object')
```

Lets look at the data types:

```
In [ ]: df.dtypes
```

```
Out[ ]: DISTRICT          object
        UPAZILA          object
        STATION_ID        object
        STATION_NAME      object
        DATE              object
        RAIN_FALL(mm)    float64
        LATITUDE          float64
        LONGITUDE         float64
        WATER_LEVEL(m)   float64
        dtype: object
```

- **DATE** is rendered as object! Need to convert it to datetime feature!

```
In [ ]: df['DATE'] = pd.to_datetime(df['DATE'])
```

```
In [ ]: df.dtypes
```

```
Out[ ]: DISTRICT          object
        UPAZILA          object
        STATION_ID        object
        STATION_NAME      object
        DATE              datetime64[ns]
        RAIN_FALL(mm)    float64
        LATITUDE          float64
        LONGITUDE         float64
        WATER_LEVEL(m)   float64
        dtype: object
```

Lets check for general infos!

```
In [ ]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1826 entries, 0 to 1825
Data columns (total 9 columns):
 #   Column           Non-Null Count  Dtype  
 ---  --  
 0   DISTRICT        1826 non-null    object 
 1   UPAZILA         1826 non-null    object 
 2   STATION_ID      1826 non-null    object 
 3   STATION_NAME    1826 non-null    object 
 4   DATE             1826 non-null    datetime64[ns]
 5   RAIN_FALL(mm)  1826 non-null    float64
 6   LATITUDE         1826 non-null    float64
 7   LONGITUDE        1826 non-null    float64
 8   WATER_LEVEL(m)  1826 non-null    float64
```

```
dtypes: datetime64[ns](1), float64(4), object(4)
memory usage: 128.5+ KB
```

Things to notice:

- There's no missing values in our dataset!

Lets check statistical properties of the numerical values:

```
In [ ]: df.describe()
```

```
Out[ ]:      RAIN_FALL(mm)  LATITUDE  LONGITUDE  WATER_LEVEL(m)
count          1826.00    1.83e+03   1.83e+03     1826.00
mean           10.00     2.18e+01   9.22e+01      6.81
std            26.09     6.04e-14   1.76e-12      0.97
min            0.00     2.18e+01   9.22e+01      5.86
25%            0.00     2.18e+01   9.22e+01      6.19
50%            0.00     2.18e+01   9.22e+01      6.50
75%            6.28     2.18e+01   9.22e+01      7.15
max           273.00    2.18e+01   9.22e+01     13.54
```

and for non numerical features:

```
In [ ]: df.describe(include=["object"])
```

```
Out[ ]:      DISTRICT  UPAZILA  STATION_ID  STATION_NAME
count          1826     1826     1826       1826
unique           1        1        1         1
top  Bandarban     Lama     CL317       Lama
freq           1826     1826     1826       1826
```

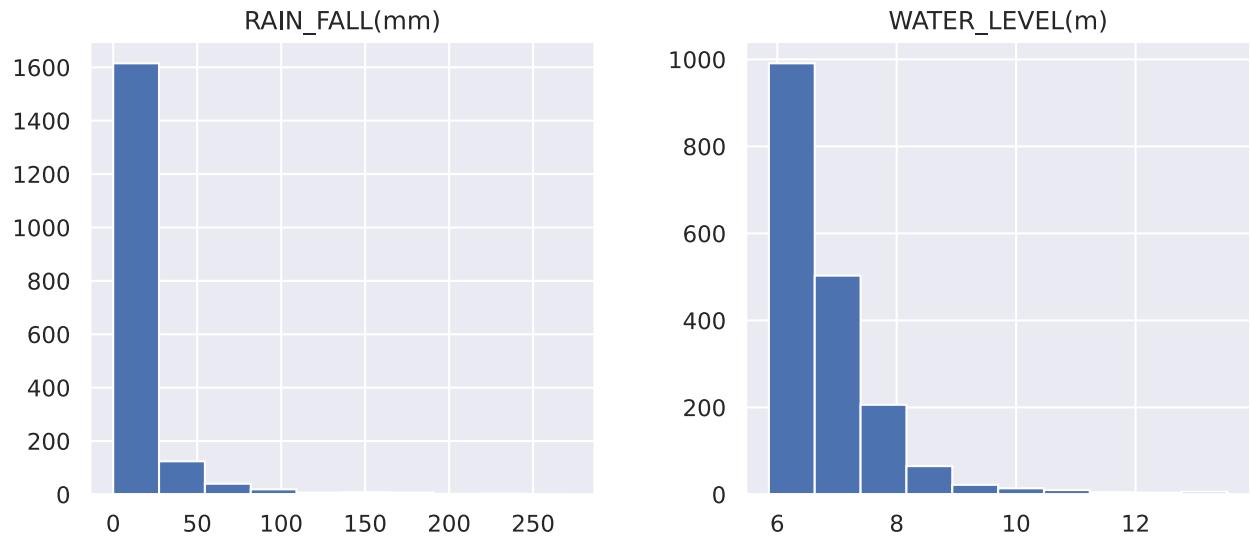
Note:

- as these categorical features only has one values, they won't contribute anything to the model!

## Visual Analysis

Histogram and Density plots

```
In [ ]: features = ['RAIN_FALL(mm)', 'WATER_LEVEL(m)']
df[features].hist(figsize=(10,4))
plt.show()
```

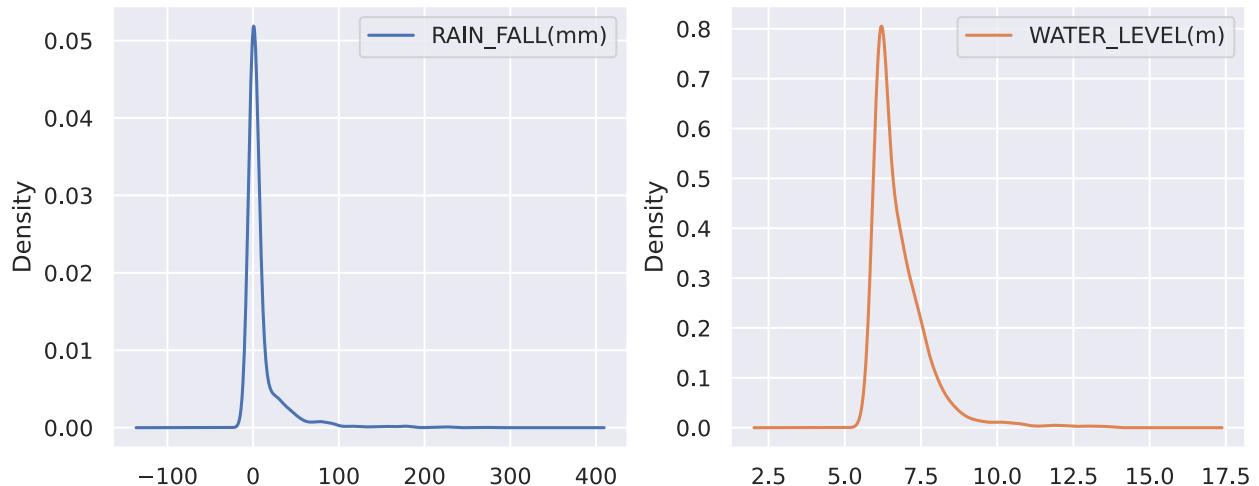


Note:

- Not normally distributed! Need to change standardize the dataset later

In [ ]:

```
df[features].plot(
    kind="density", subplots=True, layout=(1, 2), sharex=False, figsize=(10, 4)
)
plt.show()
```



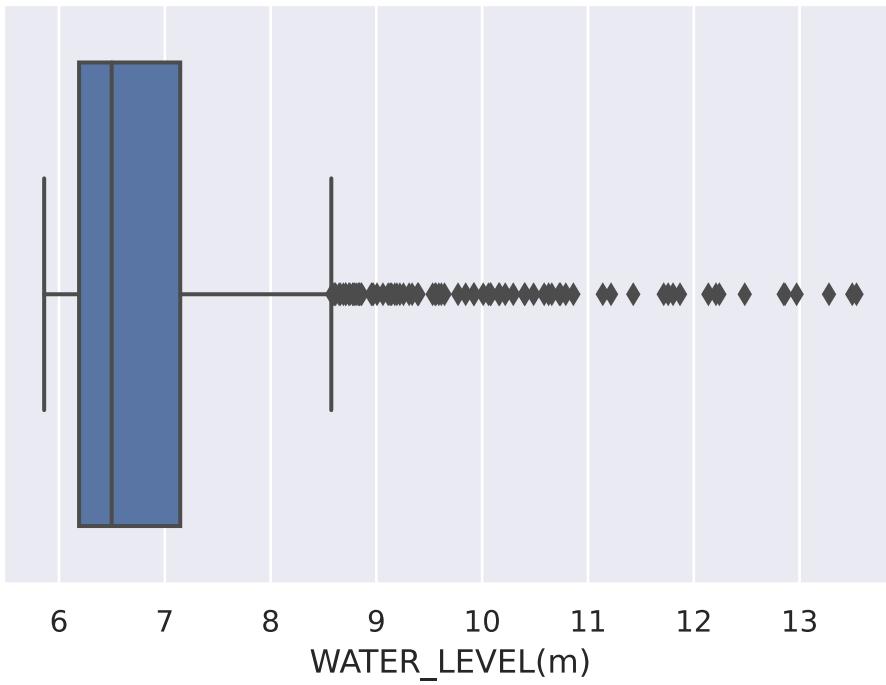
## Box Plot

Let's see how to interpret a box plot. Its components are a box (obviously, this is why it is called a box plot), the so-called whiskers, and a number of individual points (outliers).

The box by itself illustrates the interquartile spread of the distribution; its length is determined by the 25th and 75th percentiles. The vertical line inside the box marks the median (50%) of the distribution.

In [ ]:

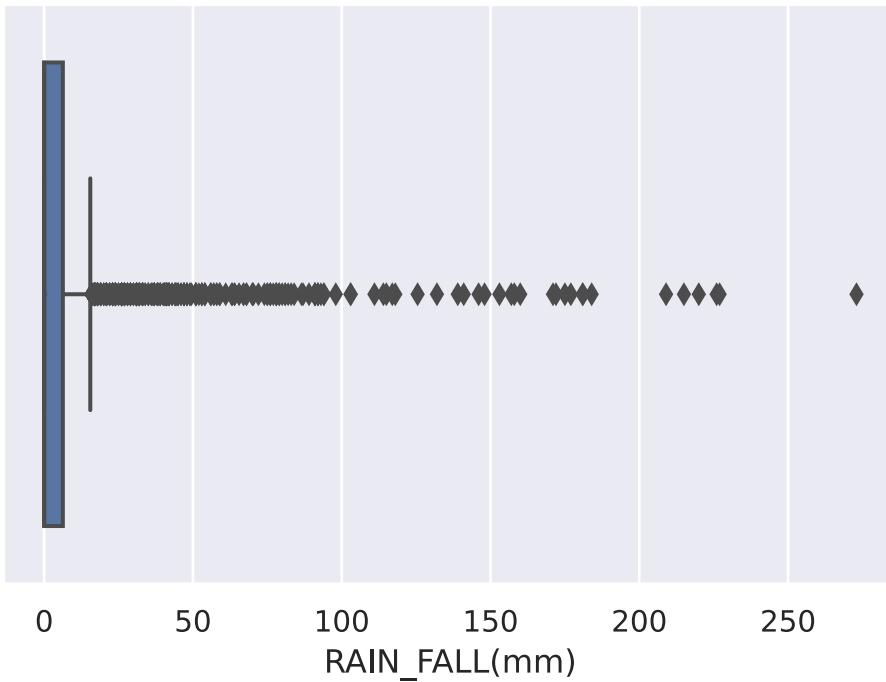
```
sns.boxplot(x="WATER_LEVEL(m)", data=df)
plt.show()
```



Note:

- There's so many outliers!

```
In [ ]: sns.boxplot(x="RAIN_FALL(mm)", data=df)
plt.show()
```



Note:

- Again so many outliers

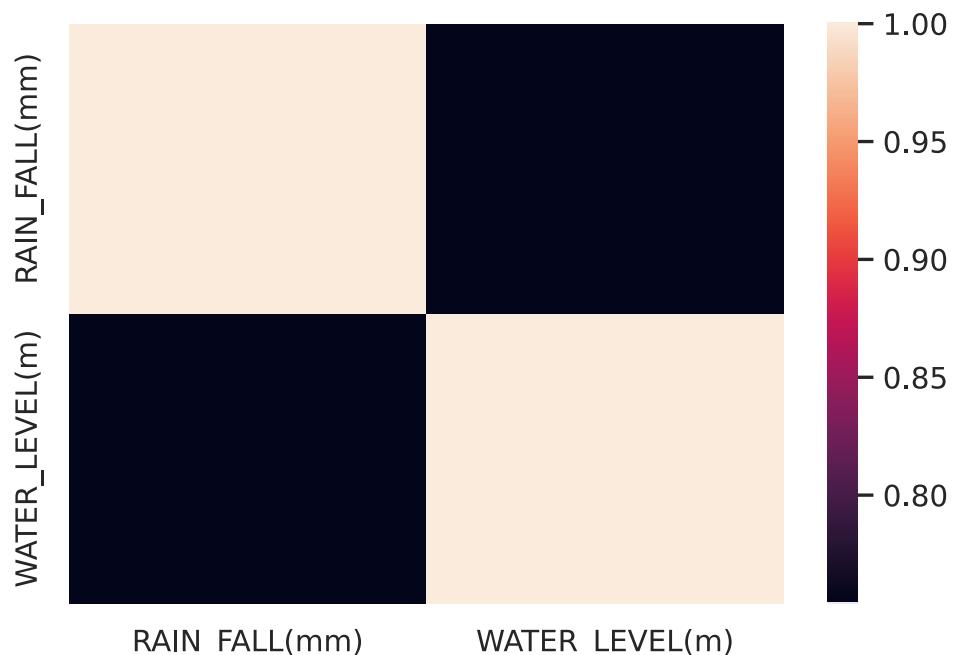
My general assumption is as this is indeed a time series data, it's intrinsic to have these outliers due to trend, seasonality and other factors regarding the weather. But as I'm not going to treat as time

series data, I have to decide whether I want these outliers to be removed or not!

## Correlation Matrix

```
In [ ]: corr_matrix = df[features].corr()  
sns.heatmap(corr_matrix)
```

```
Out[ ]: <AxesSubplot:>
```



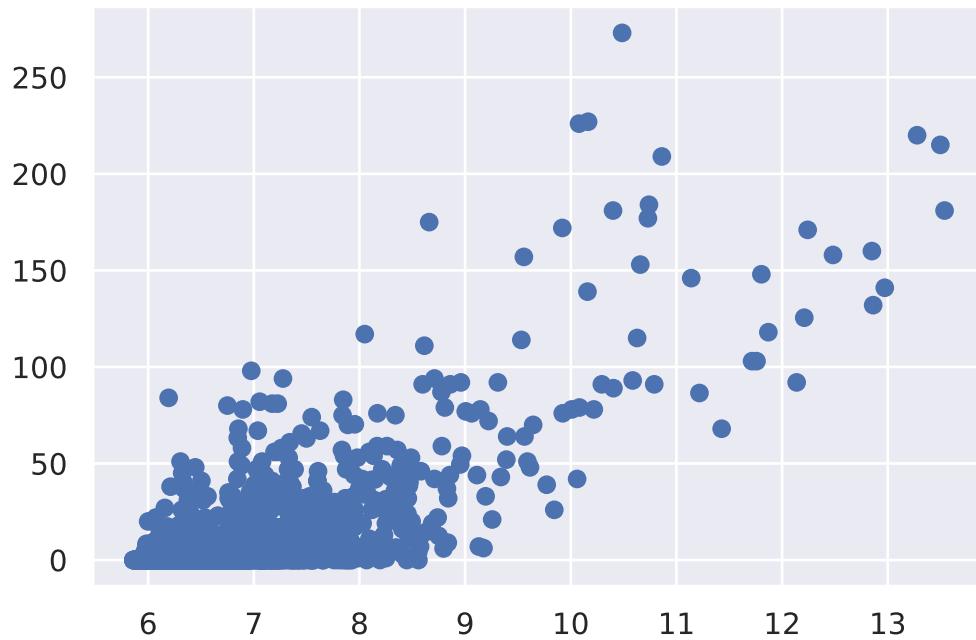
Note:

- From the visual it seems like they are around .80 correlation between rain\_fall and water\_level! (GOOD SIGN!)

## Scatter plot

The scatter plot displays values of two numerical variables as Cartesian coordinates in 2D space.

```
In [ ]: plt.scatter(df['WATER_LEVEL(m)'], df['RAIN_FALL(mm)'])  
plt.show()
```



Note:

- Not so linear due to outliers!
- But it obvious there exists positive relationship between these two features!