

```
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 are no missing values in our dataset!

Let's 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

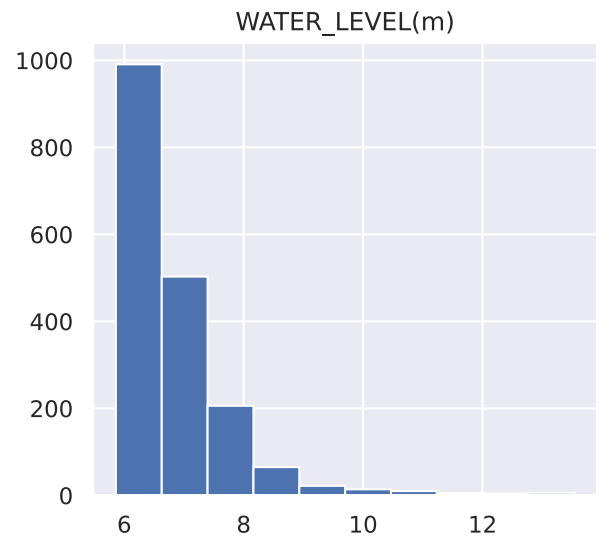
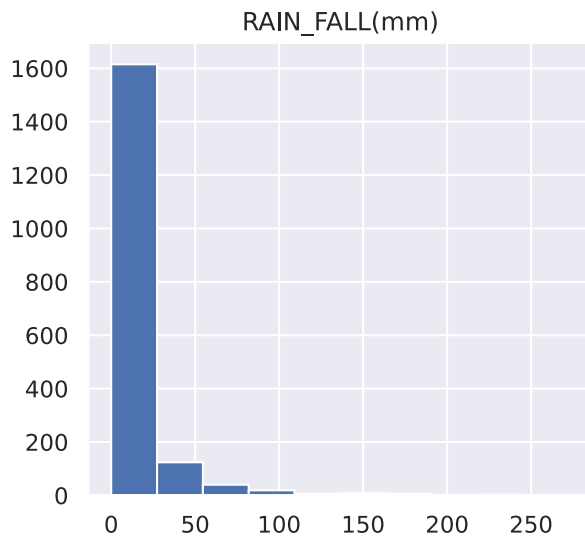
Note:

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

Visual Analysis

Histogram and Density plots

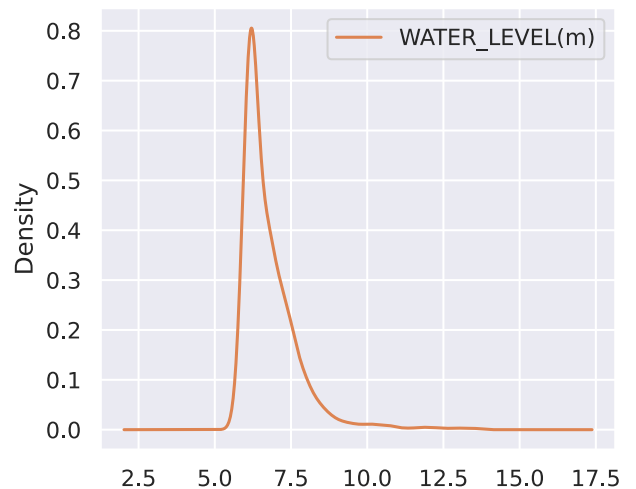
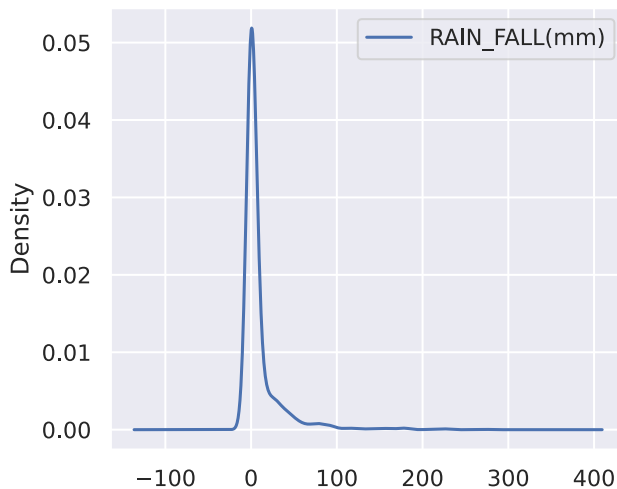
```
In [ ]: features = ['RAIN_FALL(mm)', 'WATER_LEVEL(m)']
df[features].hist(figsize=(10,4))
plt.show()
```



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()
```

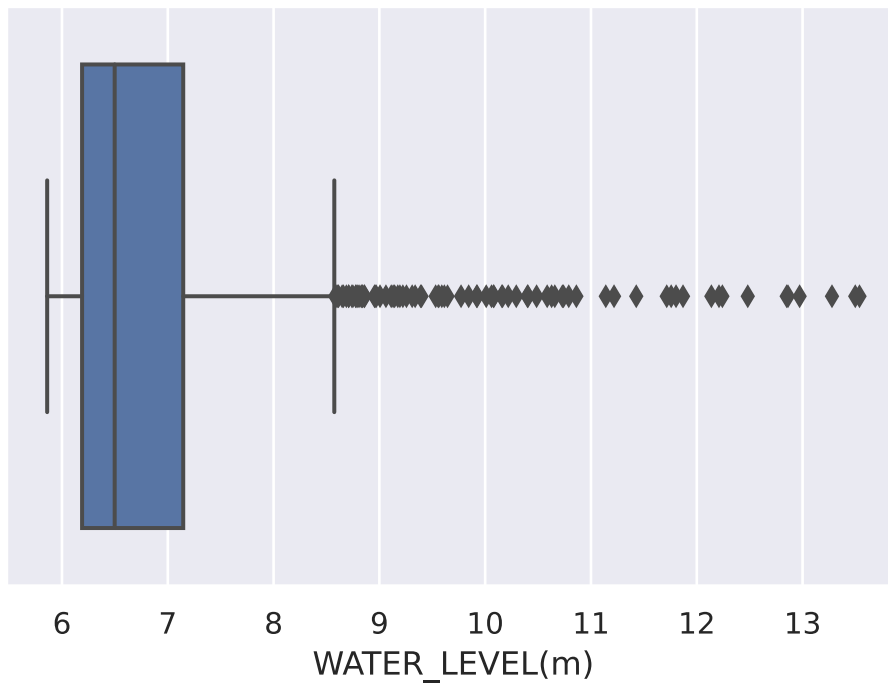


Box Plot

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.

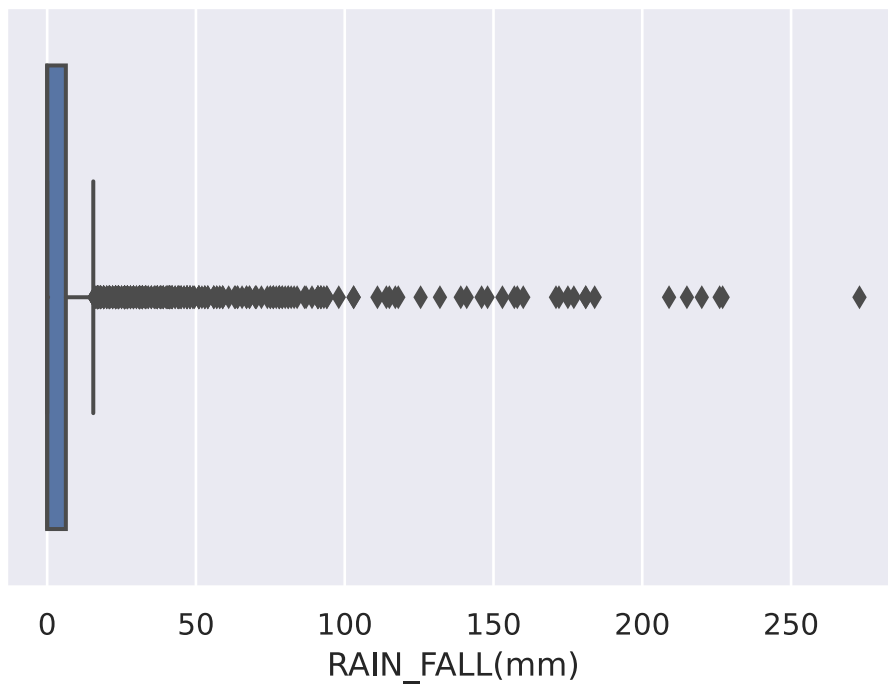
```
In [ ]: sns.boxplot(x="WATER_LEVEL(m)", data=df)
plt.show()
```



Note:

- There's so many outliers!

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



Note:

- Again so many outliers

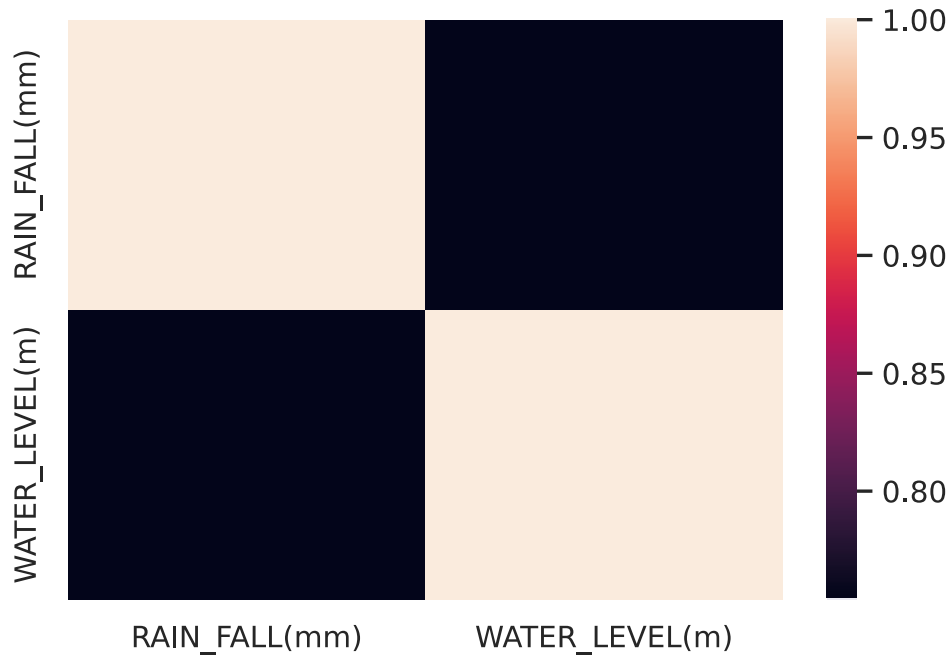
My general assumption is as this is indeed a time series data, it's intrinsic to have these outliers due to trend, seasonality and other factors regarding the weather. But as I'm not going to treat as time

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:>



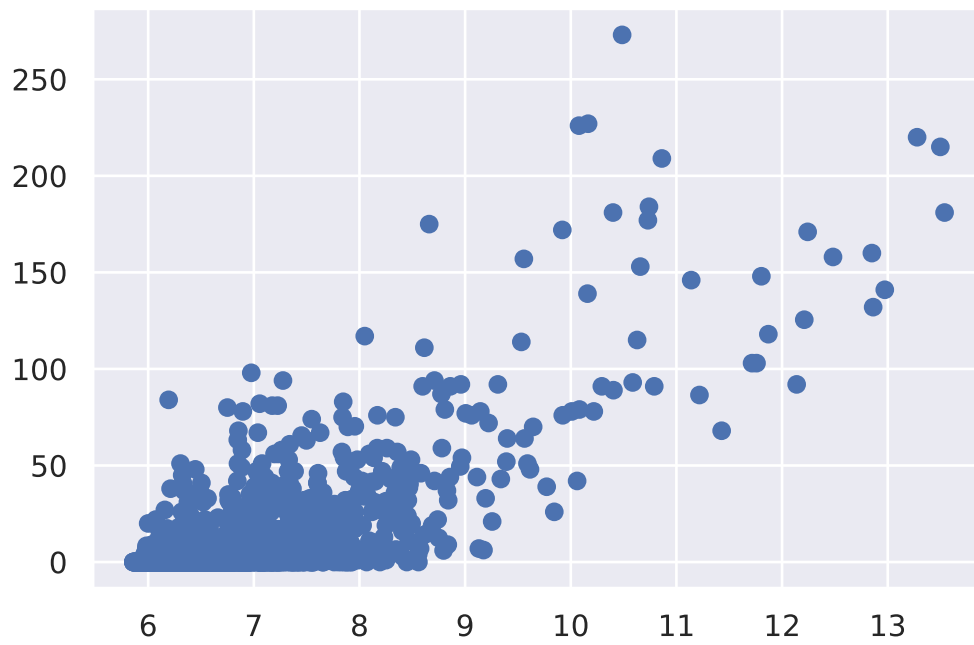
Note:

- 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!