

IS-733 DATA MINING

Homework-1

Submitted By:-

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1. Create a dataset profile table that gives an overview of the dataset.

1 Ans:- The dataset consists of data related to impurities present in the water, based on which water quality can be measured and analyzed. The dataset contains 2370 rows with 17 different attributes describing the data. By using the data profiling library, we are able to create the dataset profile table as follows;

- a.) Total number of instances/rows - **2370**
- b.) Total number of features/columns - **15**
- c.) Statistics about each column are as shown in the figure below of the dataset profile; (*Continued on the next slide...*)

Dataset Profile

Pandas Profiling Report

Overview Variables Interactions Correlations Missing values Sample

Overview

Alerts 26

Reproduction

Dataset statistics

Number of variables	15
Number of observations	2371
Missing cells	9981
Missing cells (%)	28.1%
Duplicate rows	0
Duplicate rows (%)	0.0%
Total size in memory	903.1 KiB
Average record size in memory	390.0 B

Variable types

Categorical	5
DateTime	1
Numeric	8
Text	1

The statistics for individual attributes are displayed as follows in dataset profile, giving information regarding whether it is numeric or discrete (Categorical), or temporal ;

Variables

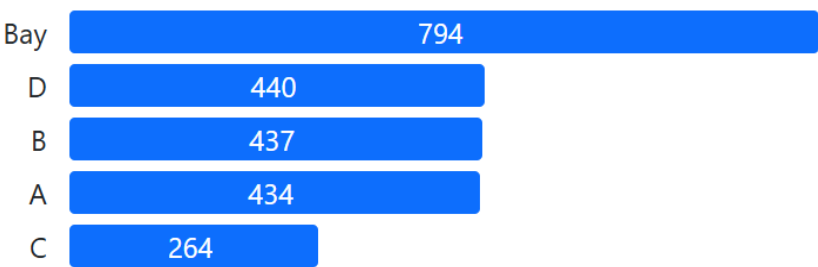
Select Columns

▼

Siteld

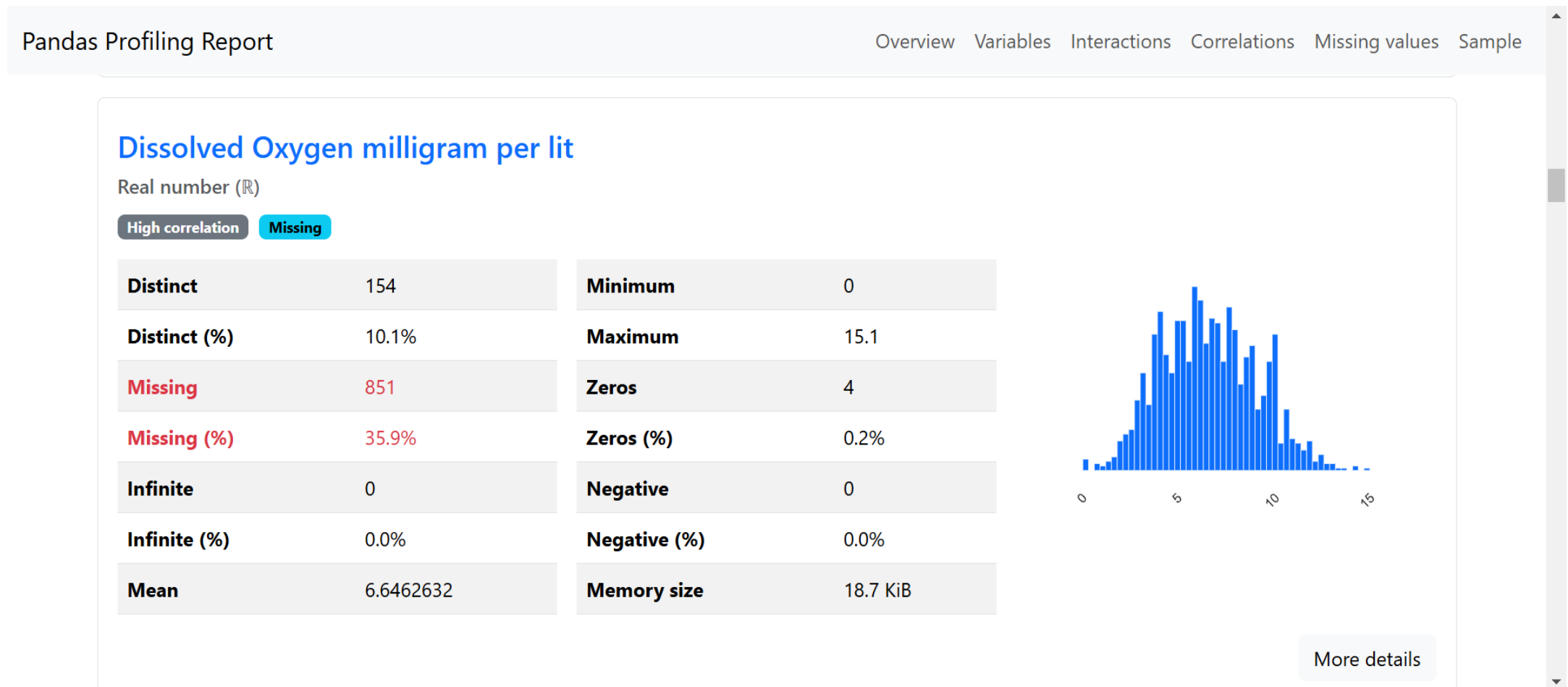
Categorical

Distinct	6
Distinct (%)	0.3%
Missing	1
Missing (%)	< 0.1%
Memory size	117.5 KiB



More details

For all the attributes, it calculates the number of distinct values, missing values, and what percentage are missing



For the numeric attributes, the values of mean, median, and standard deviation are also calculated by using the profiling package (for example, below one shows statistics of pH attribute);

Pandas Profiling Report

Overview

Variables

Interactions

Correlations

Missing values

Sample

Statistics

Histogram

Common values

Extreme values

Quantile statistics

Minimum	0.3
5-th percentile	6.5
Q1	6.5
median	7
Q3	7.5
95-th percentile	8.7
Maximum	9.9
Range	9.6
Interquartile range (IQR)	1

Descriptive statistics

Standard deviation	0.78848516
Coefficient of variation (CV)	0.10999747
Kurtosis	4.5160324
Mean	7.1682118
Median Absolute Deviation (MAD)	0.5
Skewness	0.26965677
Sum	16314.85
Variance	0.62170884
Monotonicity	Not monotonic

For the discrete attributes, we calculate the number of unique values per attribute, and top three attributes with largest count;

```
Total Number of Unique Values per Attribute:
```

```
SiteId: 6 unique values
```

```
UnitId: 2 unique values
```

```
ReadDate: 801 unique values
```

```
Time 24 clcok: 90 unique values
```

```
Field Technician: 14 unique values
```

```
DateVerified: 44 unique values
```

```
WhoVerified: 12 unique values
```

```
Top Three Attribute Values with the Largest Count:
```

```
SiteId:
```

```
SiteId
```

```
Bay    794
```

```
D      440
```

```
B      437
```

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

For each attribute, number of missing values in each column and what is the percentage of missing values is also calculated;

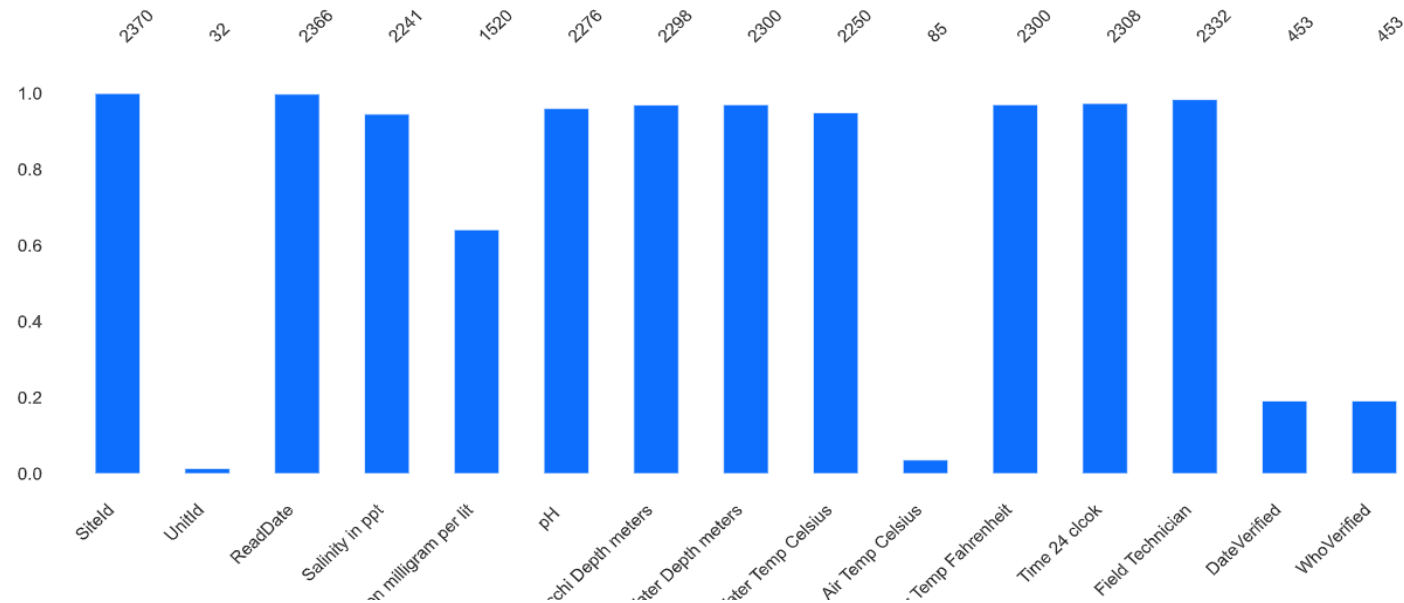
Pandas Profiling Report

[Overview](#) [Variables](#) [Interactions](#) [Correlations](#) [Missing values](#) [Sample](#)

Count

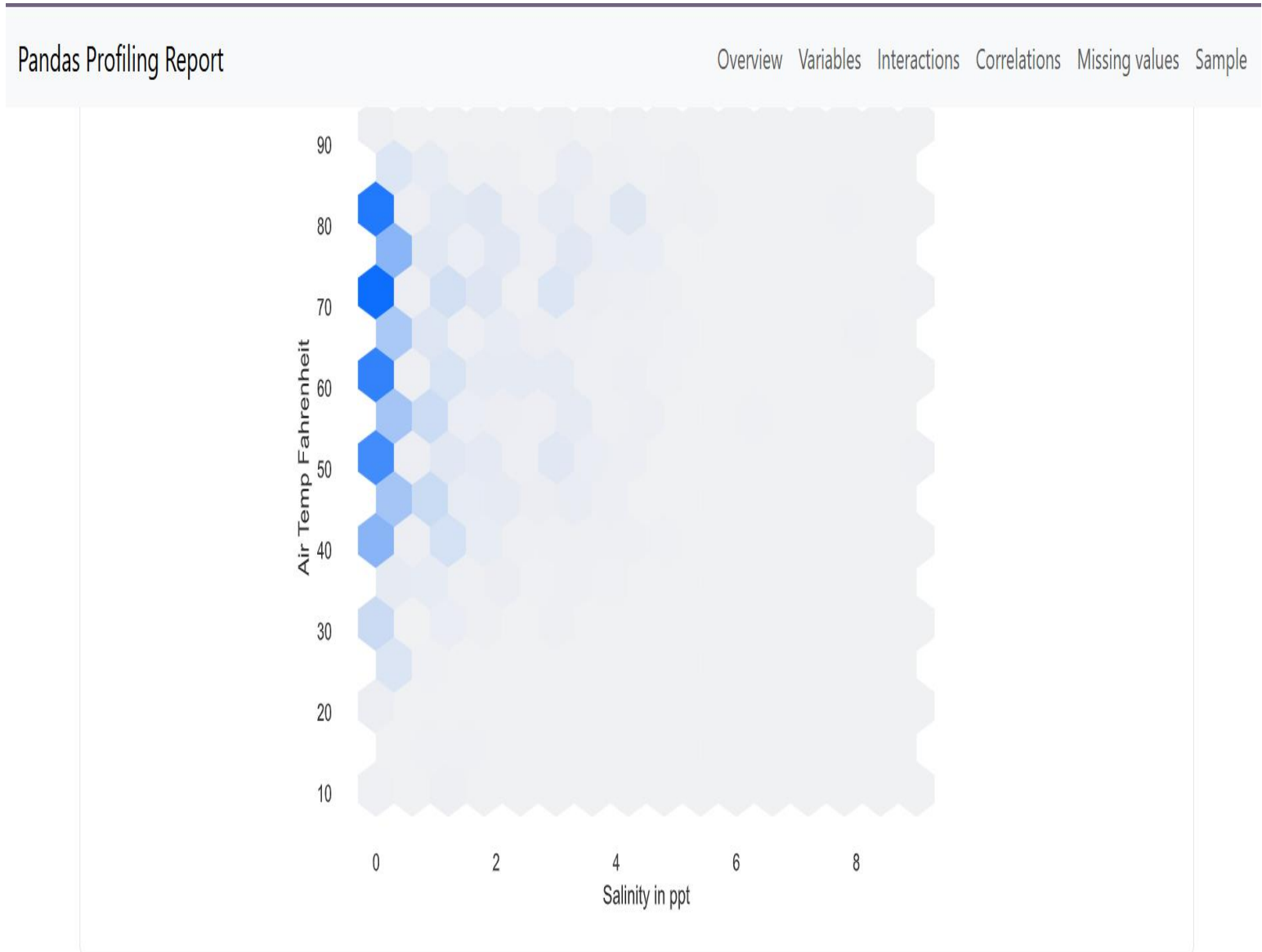
Matrix

Heatmap



- We even have the interaction plots and correlation plots, to understand the behavior of one attribute with respect to other. This will allow to understand how the attributes are affecting each other.
- We can display the sample data as well, i.e., the first and last few rows, to let user check and relate it to the data as well.

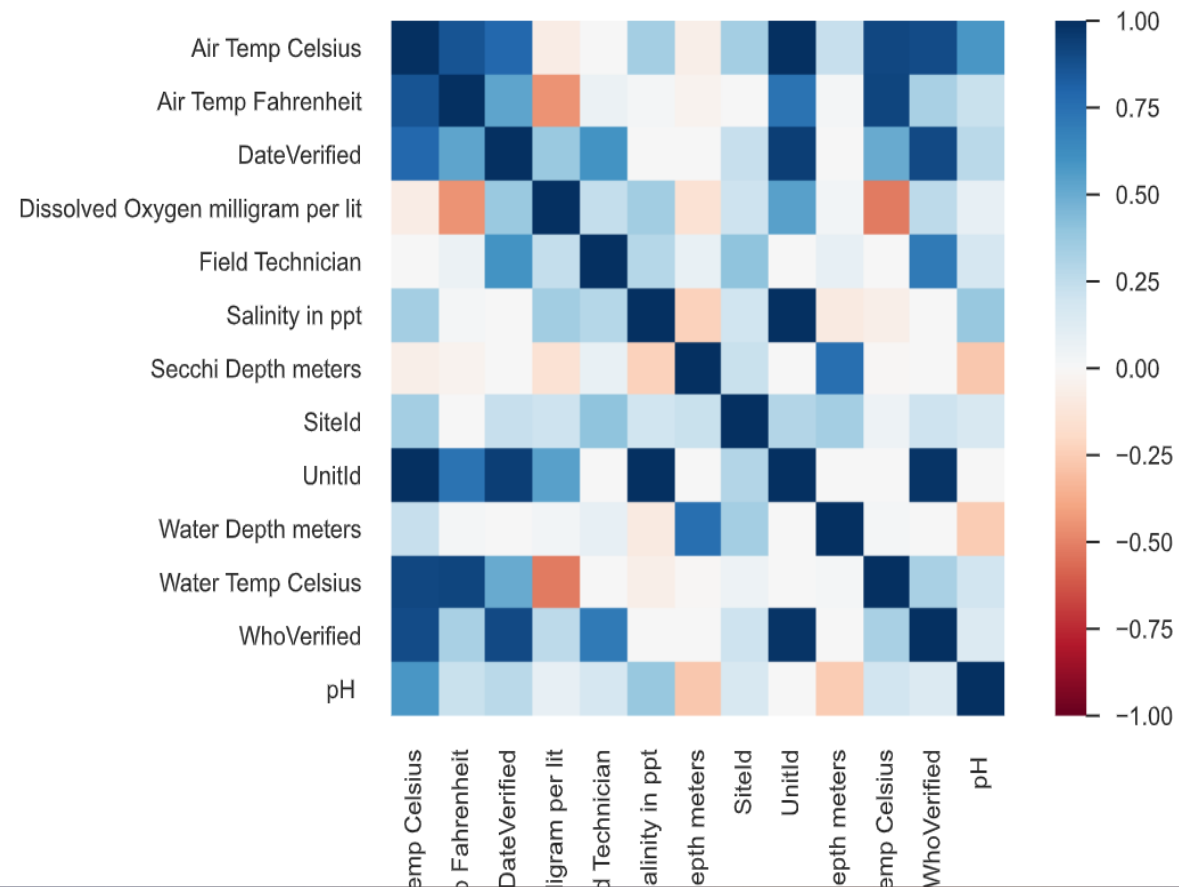
- Interactions Plot-



Correlations Plot -

Pandas Profiling Report

Overview Variables Interactions Correlations Missing values Sample



Sample data displayed in the Dataset Profiling:-

Pandas Profiling Report

[Overview](#) [Variables](#) [Interactions](#) [Correlations](#) [Missing values](#) [Sample](#)

Sample

First rows

[Last rows](#)

	Siteld	UnitId	ReadDate	Salinity in ppt	Dissolved Oxygen milligram per lit	pH	Secchi Depth meters	Water Depth meters	Water T
0	Bay	NaN	01-03-1994	1.3	11.7	7.3	0.40	0.40	5.9
1	Bay	NaN	1/31/1994	1.5	12.0	7.4	0.20	0.35	3.0
2	Bay	NaN	02-07-1994	1.0	10.5	7.2	0.25	0.60	5.9
3	Bay	NaN	2/23/1994	1.0	10.1	7.4	0.35	0.50	10.0
4	Bay	NaN	2/28/1994	1.0	12.6	7.2	0.20	0.40	1.6
5	Bay	NaN	03-07-1994	1.0	9.9	7.1	0.20	0.90	9.7
6	Bay	NaN	3/14/1994	0.5	10.4	7.2	0.25	0.75	9.8

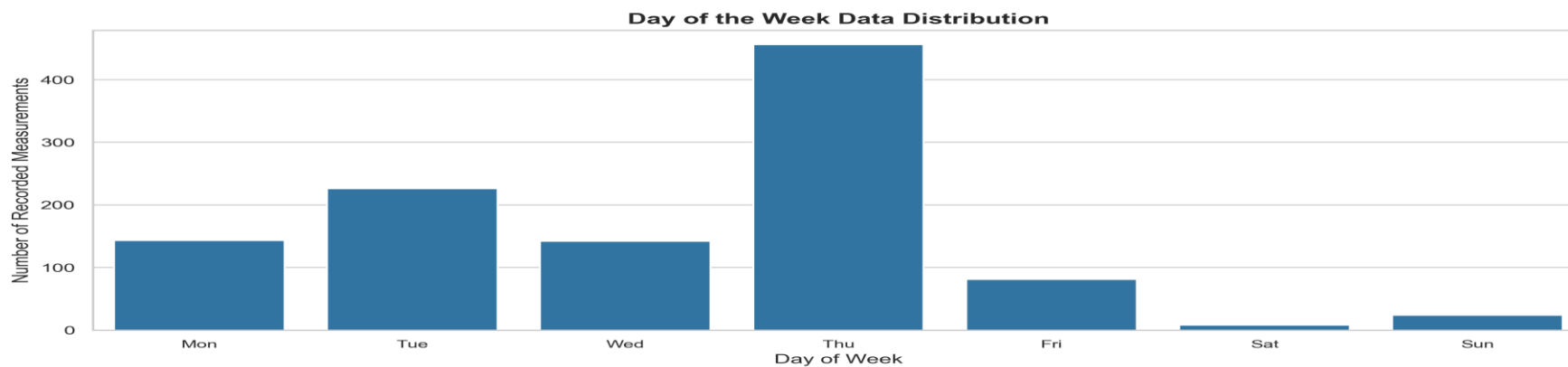
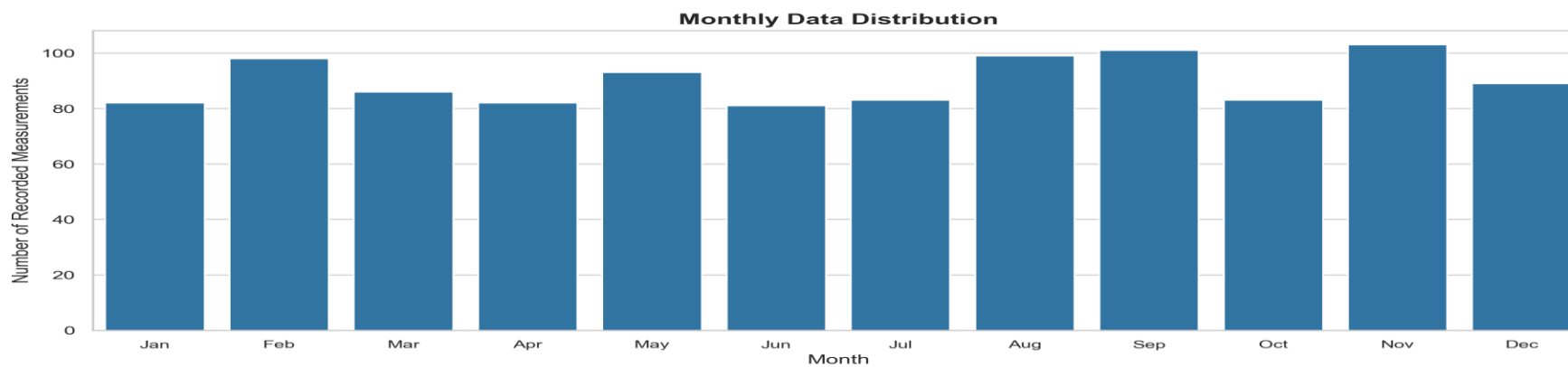
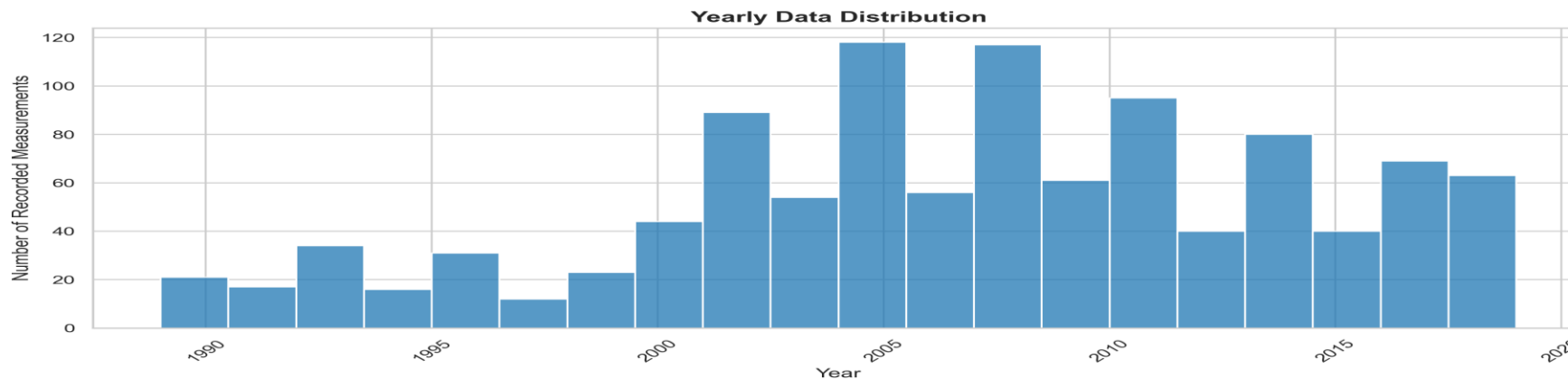
2. Generate a series of plots to describe the temporal pattern (year-to-year, monthly, and day-of-week) or other aggregate patterns.

2 Ans:- Temporal patterns generally refers to sequences of events, or trends over time. They give insights into how the data behaves in a time-dependent context. As, our dataset consists of data of water metrics over years, we can draw various plots over years, months, and day of weeks.

The temporal patterns are as shown in the below diagram;

(Continued on the next slide..)

TEMPORAL PATTERNS OF DATA



Analysis from the Temporal patterns:-

- From the yearly pattern, there is no much historical data before 1995, but the data is collected in a scheduled pattern from the years about 2000. And the maximum data is collected between the years 2005 to 2010.
- In case on the monthly trends, we could see that data is collected in more numbers during the starting of the year in January month and during the end of the year in November.
- Coming to the case of days of the week, the data collected on Thursday is highest compared to all the other days of week, and very low on weekends.

3. Generate a plot describing the distribution of your data, think of what machine learning problem could be around.

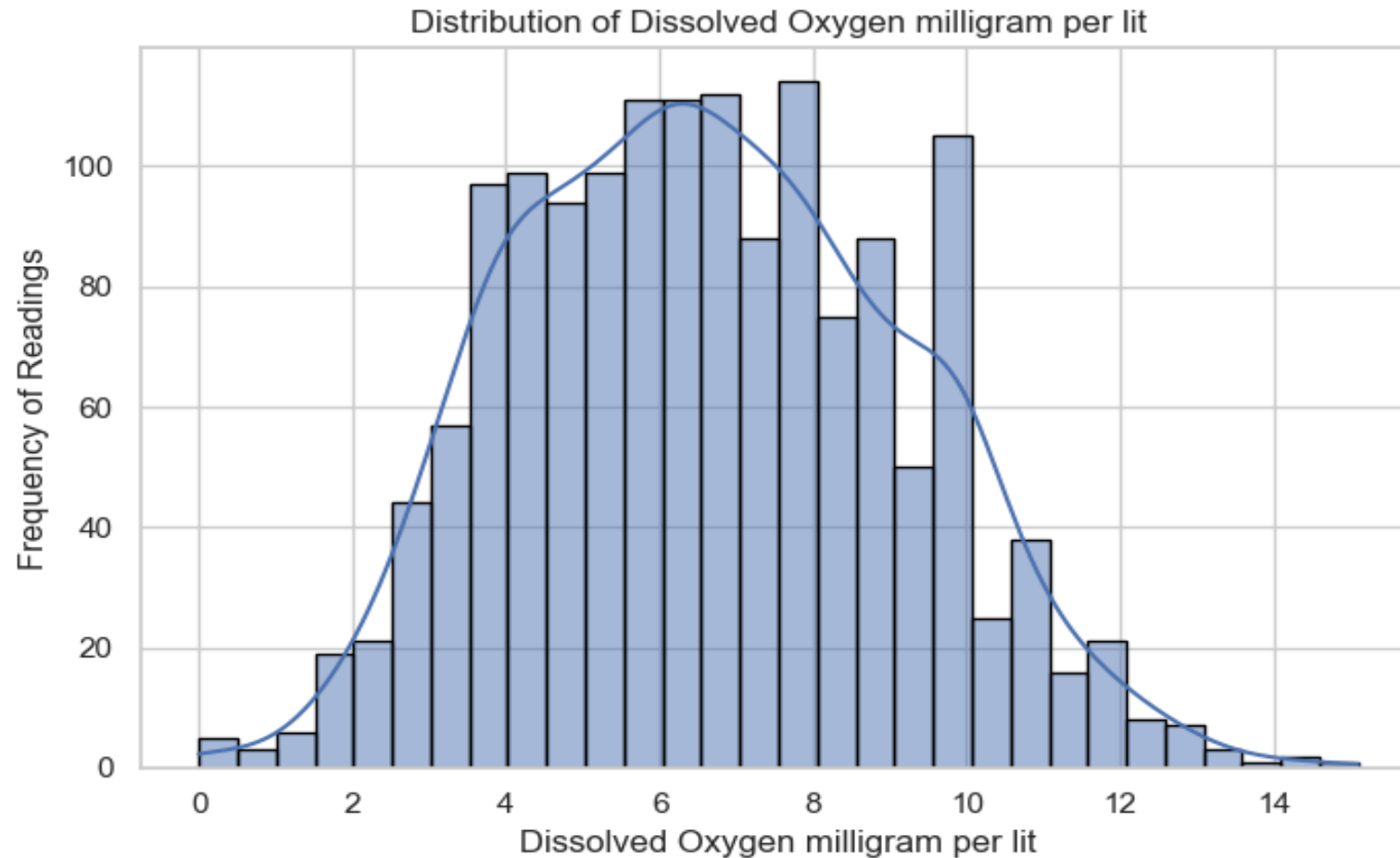
3 Ans:- The histogram plots below give the distribution of various water quality metrics. On the y-axis the frequency of the readings is taken and on the x-axis the water quality metric. Based on the distribution of data and after analyzing the data, one machine learning problem would be **Analyzing whether the water quality is good or bad.**

Using the metrics, we can draw insights whether the water is good or bad and can conclude whether it can be used or not.

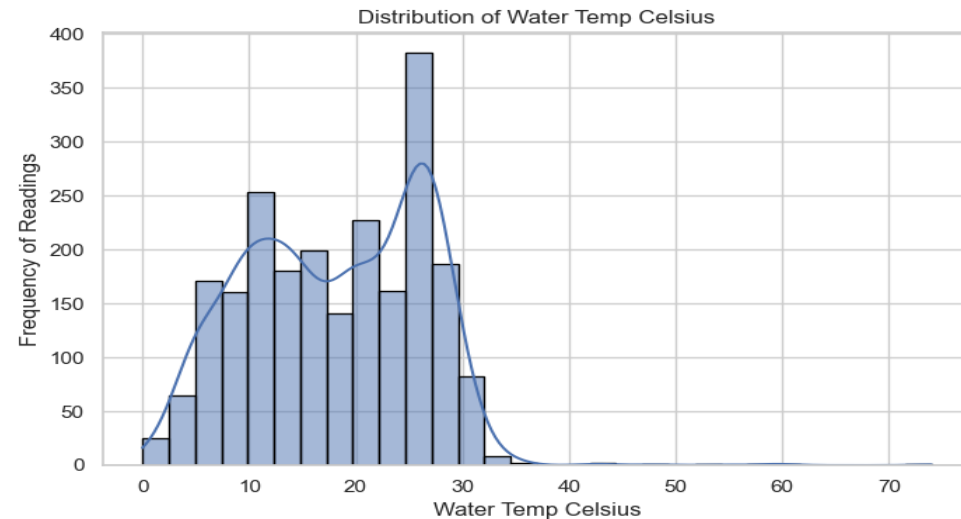
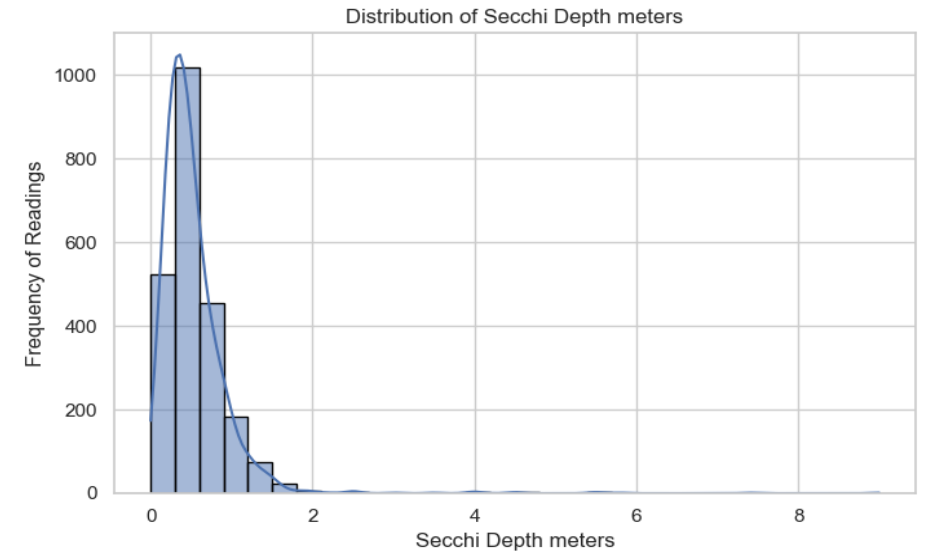
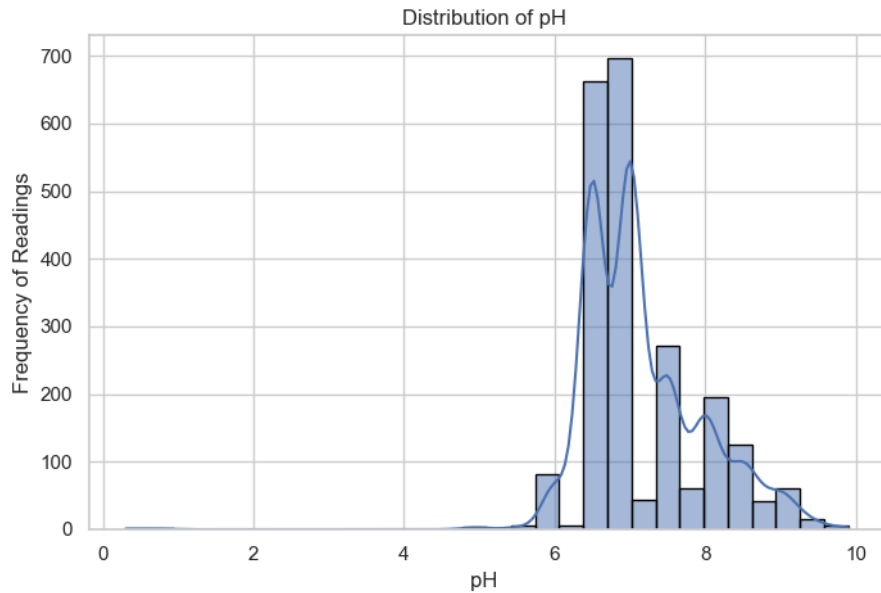
(Continued on next slide ..)

Distribution of Data for various Metrics:-

(Below is the plot for Dissolved oxygen)



Similarly, the plots for pH, Secchi Depth, Water temperature are as follows;



The basic description and information regarding the metrics is as follows;

```
RangeIndex: 2371 entries, 0 to 2370  
Data columns (total 15 columns):
```

#	Column	Non-Null Count	Dtype
0	SiteId	2370 non-null	object
1	UnitId	32 non-null	object
2	ReadDate	2366 non-null	object
3	Salinity in ppt	2241 non-null	float64
4	Dissolved Oxygen milligram per lit	1520 non-null	float64
5	pH	2276 non-null	float64
6	Secchi Depth meters	2298 non-null	float64
7	Water Depth meters	2300 non-null	float64
8	Water Temp Celsius	2250 non-null	float64
9	Air Temp Celsius	85 non-null	float64
10	Air Temp Fahrenheit	2300 non-null	float64
11	Time 24 clcok	2308 non-null	object
12	Field Technician	2332 non-null	object
13	DateVerified	453 non-null	object
14	WhoVerified	453 non-null	object

```
dtypes: float64(8), object(7)  
memory usage: 278.0+ KB  
None
```

	Salinity in ppt	Dissolved Oxygen milligram per lit	pH	\
count	2241.000000	1520.000000	2276.000000	
mean	0.717068	6.646263	7.168212	
std	1.230819	2.506608	0.788485	
min	0.000000	0.000000	0.300000	
25%	0.000000	4.800000	6.500000	
50%	0.000000	6.500000	7.000000	
75%	1.000000	8.500000	7.500000	
max	9.000000	15.100000	9.900000	

	Secchi Depth meters	Water Depth meters	Water Temp Celsius	\
count	2298.000000	2300.000000	2250.000000	
mean	0.524898	0.762559	18.062138	
std	0.473663	0.621140	8.298246	
min	0.000000	0.010000	0.000000	
25%	0.300000	0.400000	11.000000	
50%	0.400000	0.650000	19.000000	
75%	0.650000	0.950000	25.000000	
max	9.000000	12.000000	74.000000	

	Air Temp Celsius	Air Temp Fahrenheit
count	85.000000	2300.000000
mean	16.437647	62.051637
std	11.754138	15.492236
min	0.000000	10.500000
25%	9.000000	49.000000
50%	15.000000	63.000000
75%	21.700000	75.000000
max	74.000000	92.300000

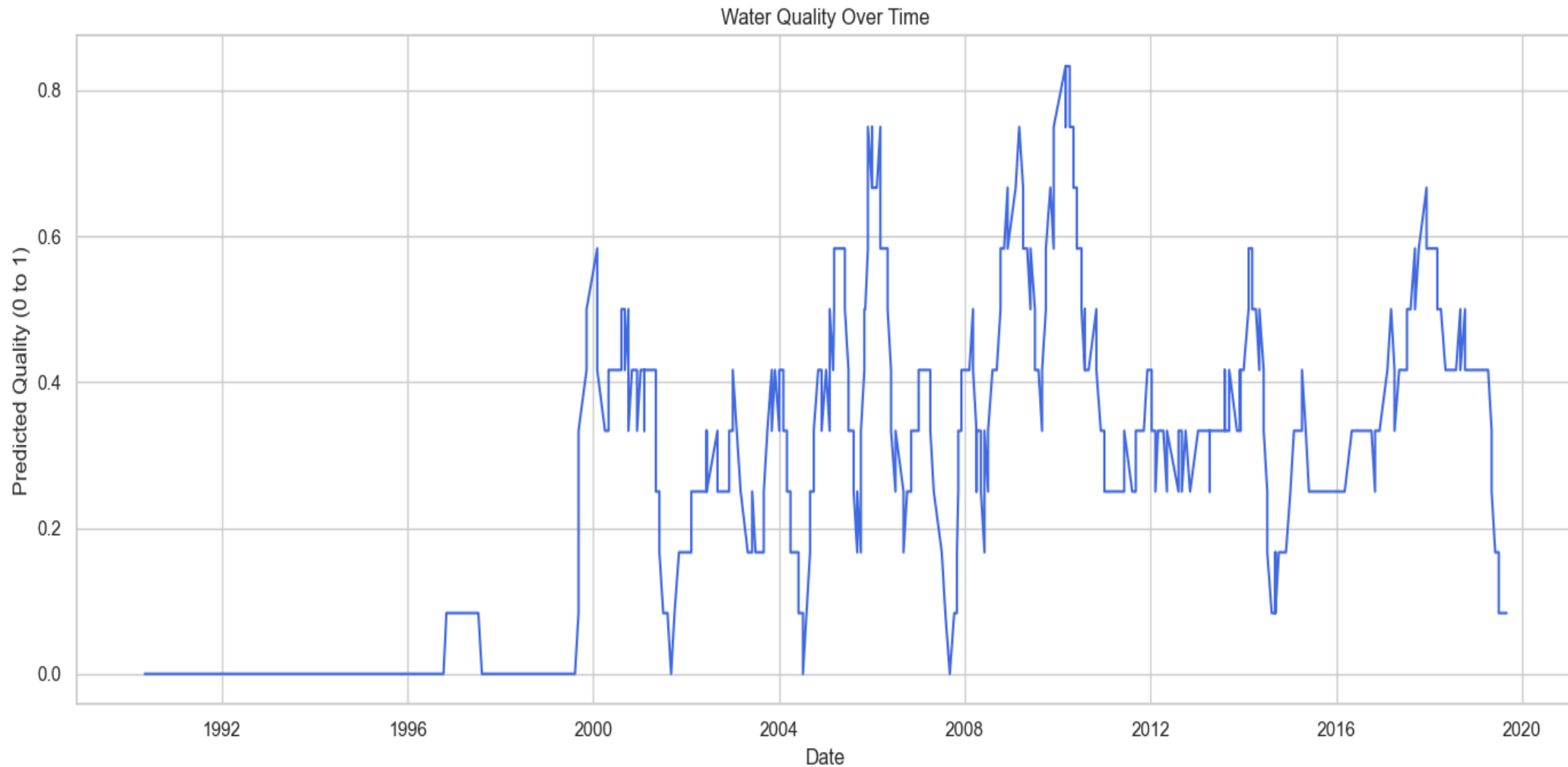
4. Generate a series of plots to illustrate to support your story and make your points clear.

4 Ans:- To show how the water quality is predicted, whether it is good or bad, we can use ML techniques like Random forest to analyze the data and train the model (We are using 80% of the data from the dataset to train the model, and 20% of the data to test the model). Once the model, is trained on the data, it will be able to determine whether water quality is good or bad.

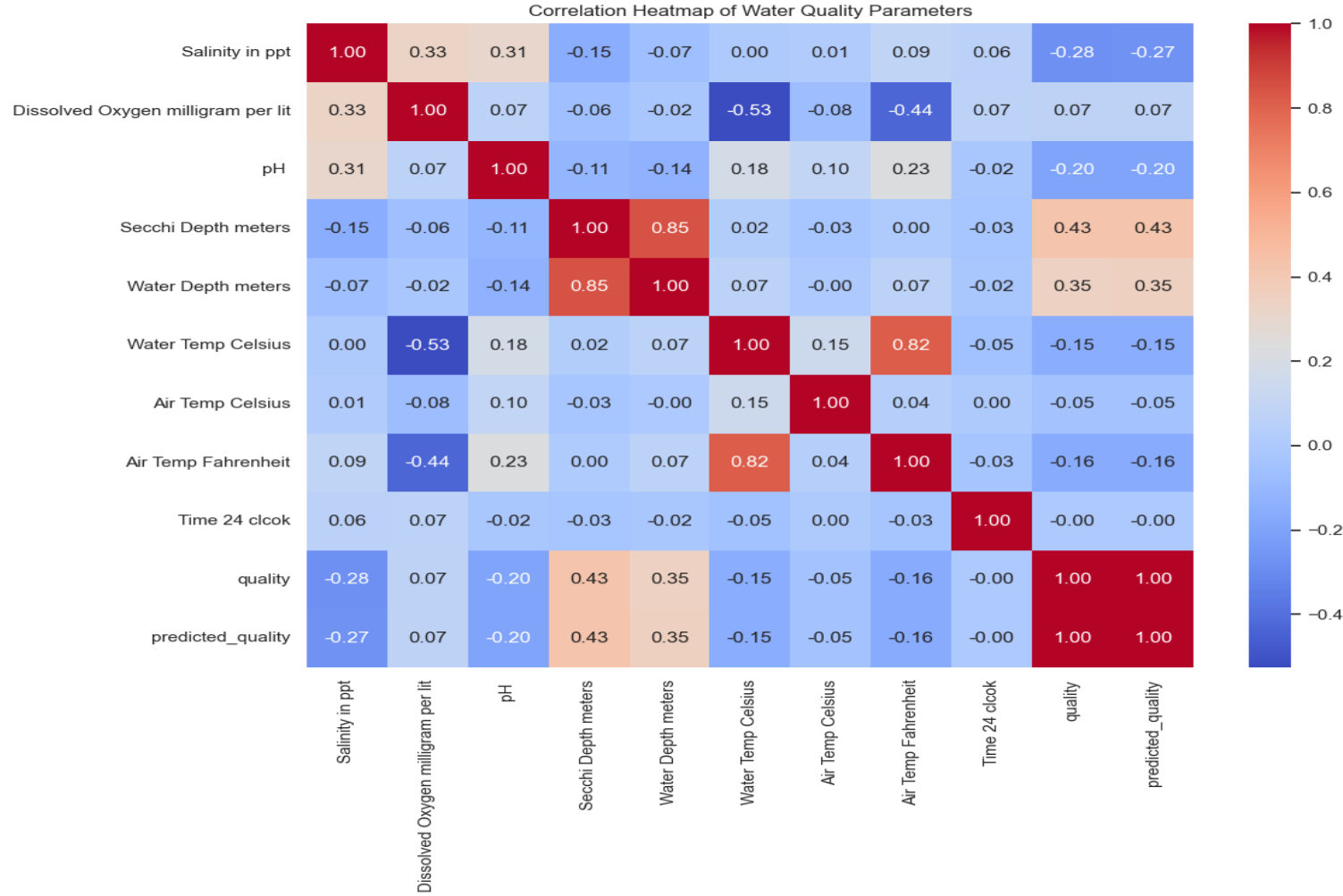
By using the line plot and heat map we understand the water quality. Please find the plots on the next slides to understand the pattern in the data;

(Continued on next slides..)

Line Plot of Water Quality



Water Quality Correlation Heat Map



Insights from Line plot and Heatmap:-

- The line plot clearly gives the statistics over the years on how the water quality is changing. Based on the metrics, water quality is measured (i.e. **0** - indicates **Bad or Low Quality**, **1** – indicates **Good or High Quality**). During the years, between **2008** and **2012** the water quality improved a lot and was good till **2011**, later it got reduced and gradually increased by **2018**.
- The correlation Heatmap, gives the information regarding which **parameters must be increased**, and **which must be decreased**. The **one's with negative values must be reduced** to maximum extent to increase the water quality and **one's with positive must be increased** to maintain or increase the quality of water.
- From the heatmap plot, **pH, salinity in water, water temp., air temperature must be reduced**, and **Dissolved oxygen, Secchi depth, and water depth must be increased** to increase water quality.

5. Design a dashboard that allows users to explore the data pattern. You may get inspiration from tasks 2-4, but feel free to add insights.

5 Ans:- A Dashboard is created with all the water quality metrics spread over years span, allowing users to explore different patterns of water quality over years. Provided are the options to zoom in and out, download the plot in case if required, reset the axes, Auto scale the plot.

There are options to even highlight the plot to have a closer look at the value at that time instance. This way we can compare overall span with one particular date as required. Below is the dashboard generated;

(Continued on next slide..)

Dashboard showing Data Insights

Water Quality Dashboard

1989-05-11 → 2019-11-05

pH

×

pH Over Time

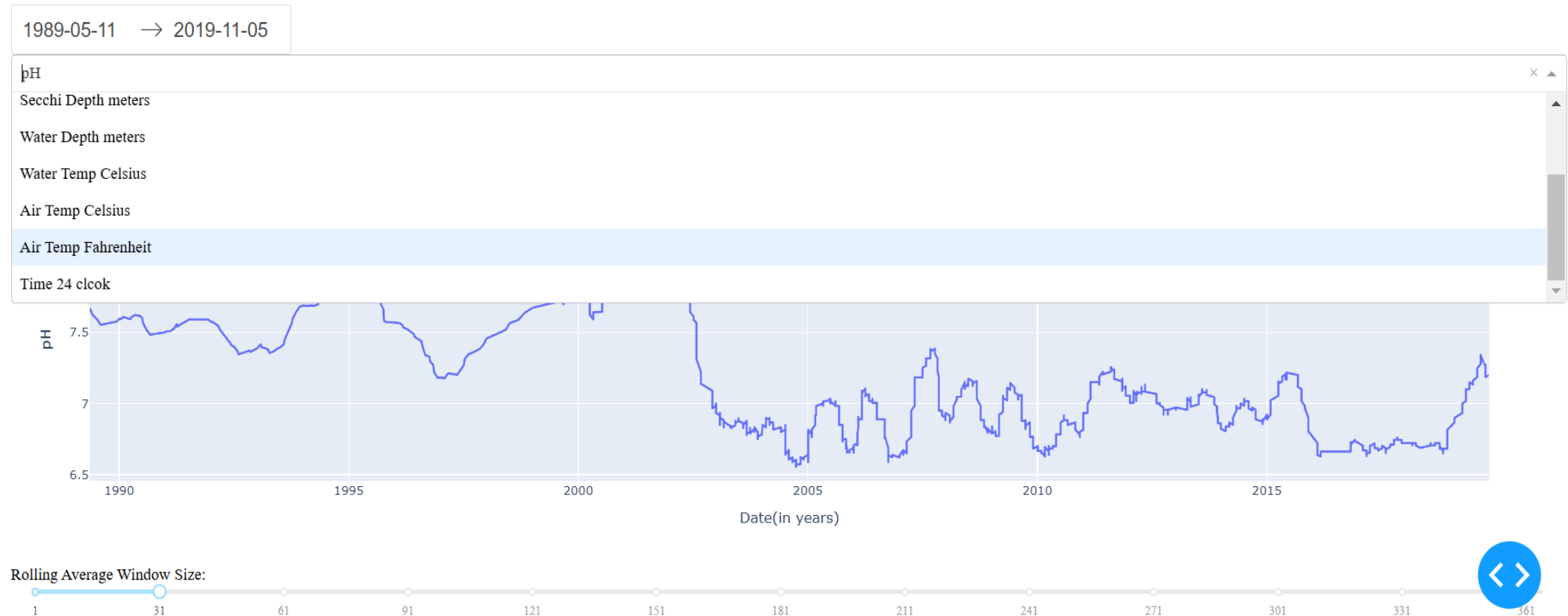


Rolling Average Window Size:



The Dashboard provides dropdown as well to select various attributes and check their trends over years;

Water Quality Dashboard



There is even a scaler as "**Rolling average window size**", which is used to smooth out fluctuations during the small intervals and reduce the noise. This helps to adjust size according the requirement, recent changes(smaller window) and overall (larger window).



Using of the Dashboard :-

- To view the dashboard, it is necessary for the dash application to run and generate the plot.
- So, need to execute code related to Question-5 to get the server running and dash application running, to generate the dashboard.
- The Dashboard is found at the following address:-

<http://127.0.0.1:8050/>

(Please run the code before clicking on the link to view the dashboard)

Conclusion

- The water quality is determined over years, and over the years between **2008 to 2011 it has greatly improved**, and after that it gradually decreased and maintained steady fluctuations at **0.6-0.8** range (i.e., where water quality is Average) over years of **2018 to 2020**.
- The dashboard gives clear understanding on how the attributes are changing over the period of time. Especially Salinity, pH, and Dissolved oxygen which are very much important to determine the water quality.
- One more important point that is understood from the plots is, there is need to increase the dissolved oxygen in water to make it good and decrease the pH, and salinity of the water.

The Corresponding python code is uploaded to GitHub. Please find it in the below link:-

https://github.com/UB01976/is7332025/blob/main/data-mining-project-repo/UB01976_02242025_HW1

The folder consists of the Dataset, Python codes file to generate the plots, Dataset Profile report, and jupyter notebook.

References:-

- Dataset - <https://catalog.data.gov/dataset/water-quality-data-41c5e>
- Dataset Landing page - <https://iris.fws.gov/APPS/ServCat/Reference/Profile/117348>