

PHASE:3

Water Quality Analysis

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Introduction:

Water quality analysis is a critical aspect of environmental science and public health, aiming to assess the safety and health of water sources for various purposes, such as drinking, agriculture, industrial use, and aquatic ecosystems. Data analytics plays a vital role in this field by enabling the collection, processing, interpretation, and visualization of data related to water quality. It helps researchers, environmentalists, and policymakers make informed decisions about managing water resources and protecting public health.

Data Collection:

- Water quality data is collected from various sources, including rivers, lakes, reservoirs, groundwater, and treatment facilities.
- Data may include measurements of physical, chemical, and biological parameters, such as temperature, pH, turbidity, dissolved oxygen, nutrients, heavy metals, and microbial contaminants.
- Sensors, monitoring stations, and sampling methods are used to collect data over time, providing a detailed picture of water quality dynamics.

DatasetLink: <https://www.kaggle.com/datasets/adityakadiwal/water-potability>

Data Preprocessing and Cleaning:

- Clean the collected data to ensure its quality and accuracy.

#importing data set

import pandas as pd

import numpy as np

```
import matplotlib.pyplot as plt

import seaborn as sns

main_dat = pd.read_csv("water_potability.csv")

ks = main_dat.copy() #copy of original data set

ks.head()
```

OUTPUT:

SECTION-1 (Data preprocessing)

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[2]: main_dat = pd.read_csv("water_potability.csv")
ks = main_dat.copy() #copy of original data set
```

```
[3]: ks.head()
```

```
[3]:
```

	ph	Hardness	Solids	Chloramines	Sulfate	Conductivity	Organic_carbon	Trihalomethanes	Turbidity	Potability
0	NaN	204.890455	20791.318981	7.300212	368.516441	564.308654	10.379783	86.990970	2.963135	0
1	3.716080	129.422921	18630.057858	6.635246	NaN	592.885359	15.180013	56.329076	4.500656	0
2	8.099124	224.236259	19909.541732	9.275884	NaN	418.606213	16.868637	66.420093	3.055934	0
3	8.316766	214.373394	22018.417441	8.059332	356.886136	363.266516	18.436524	100.341674	4.628771	0
4	9.092223	181.101509	17978.986339	6.546600	310.135738	398.410813	11.558279	31.997993	4.075075	0

ks.sample(5)

ks.shape

ks.columns

OUTPUT:

```
[4]: ks.sample(5)
```

```
[4]:
```

	ph	Hardness	Solids	Chloramines	Sulfate	Conductivity	Organic_carbon	Trihalomethanes	Turbidity	Potability
1018	6.013161	218.843256	21573.747571	9.295852	321.168313	444.276635	14.744347	62.443239	3.455623	0
248	6.581878	272.982745	37169.444404	8.114731	416.083481	351.476839	15.129334	79.261026	4.201663	0
1998	7.544306	211.051146	34359.400797	8.166793	365.812313	447.520655	18.553478	60.162746	3.714096	1
2227	NaN	159.832881	23917.190146	6.781576	369.223852	472.927194	13.891834	85.758645	2.857687	0
2484	6.653650	172.584512	34816.444538	8.289307	293.611048	389.471149	15.872474	67.976869	4.871406	0

```
[5]: ks.shape
```

```
[5]: (3276, 10)
```

```
[6]: ks.columns
```

```
[6]: Index(['ph', 'Hardness', 'Solids', 'Chloramines', 'Sulfate', 'Conductivity',  
        'Organic_carbon', 'Trihalomethanes', 'Turbidity', 'Potability'],  
        dtype='object')
```

pd.isnull(ks).sum()

ks.dropna(inplace=True)

pd.isnull(ks).sum()

OUTPUT:

```
[7]: pd.isnull(ks).sum()
```

```
[7]: ph          491  
     Hardness      0  
     Solids        0  
     Chloramines   0  
     Sulfate       781  
     Conductivity  0  
     Organic_carbon 0  
     Trihalomethanes 162  
     Turbidity     0  
     Potability    0  
     dtype: int64
```

```
[8]: ks.dropna(inplace=True)  
     pd.isnull(ks).sum()
```

```
[8]: ph          0  
     Hardness      0  
     Solids        0  
     Chloramines   0  
     Sulfate       0  
     Conductivity  0  
     Organic_carbon 0  
     Trihalomethanes 0  
     Turbidity     0  
     Potability    0  
     dtype: int64
```

ks.describe()

ks.nunique()

OUTPUT:

```
[9]: ks.describe()

[9]:
```

	ph	Hardness	Solids	Chloramines	Sulfate	Conductivity	Organic_carbon	Trihalomethanes	Turbidity	Potability
count	2011.000000	2011.000000	2011.000000	2011.000000	2011.000000	2011.000000	2011.000000	2011.000000	2011.000000	2011.000000
mean	7.085990	195.968072	21917.441374	7.134338	333.224672	426.526409	14.357709	66.400859	3.969729	0.403282
std	1.573337	32.635085	8642.239815	1.584820	41.205172	80.712572	3.324959	16.077109	0.780346	0.490678
min	0.227499	73.492234	320.942611	1.390871	129.000000	201.619737	2.200000	8.577013	1.450000	0.000000
25%	6.089723	176.744938	15615.665390	6.138895	307.632511	366.680307	12.124105	55.952664	3.442915	0.000000
50%	7.027297	197.191839	20933.512750	7.143907	332.232177	423.455906	14.322019	66.542198	3.968177	0.000000
75%	8.052969	216.441070	27182.587067	8.109726	359.330555	482.373169	16.683049	77.291925	4.514175	1.000000
max	14.000000	317.338124	56488.672413	13.127000	481.030642	753.342620	27.006707	124.000000	6.494749	1.000000

```
[10]: ks.nunique()

[10]:
```

ph	2011
Hardness	2011
Solids	2011
Chloramines	2011
Sulfate	2011
Conductivity	2011
Organic_carbon	2011
Trihalomethanes	2011
Turbidity	2011
Potability	2
dtype:	int64

ks.info()

ks.dtypes

OUTPUT:

```
[11]: ks.info()

<class 'pandas.core.frame.DataFrame'>
Index: 2011 entries, 3 to 3271
Data columns (total 10 columns):
 #   Column              Non-Null Count  Dtype
---  ---
 0   ph                   2011 non-null   float64
 1   Hardness             2011 non-null   float64
 2   Solids               2011 non-null   float64
 3   Chloramines          2011 non-null   float64
 4   Sulfate              2011 non-null   float64
 5   Conductivity         2011 non-null   float64
 6   Organic_carbon       2011 non-null   float64
 7   Trihalomethanes      2011 non-null   float64
 8   Turbidity            2011 non-null   float64
 9   Potability           2011 non-null   int64
dtypes: float64(9), int64(1)
memory usage: 172.8 KB

[12]: ks.dtypes

[12]:
```

ph	float64
Hardness	float64
Solids	float64
Chloramines	float64
Sulfate	float64
Conductivity	float64
Organic_carbon	float64
Trihalomethanes	float64
Turbidity	float64
Potability	int64
dtype:	object

#finding the correlation

ks.corr()

OUTPUT:

[13]: #finding the correlation
ks.corr()

[13]:

	ph	Hardness	Solids	Chloramines	Sulfate	Conductivity	Organic_carbon	Trihalomethanes	Turbidity	Potability
ph	1.000000	0.108948	-0.087615	-0.024768	0.010524	0.014128	0.028375	0.018278	-0.035849	0.014530
Hardness	0.108948	1.000000	-0.053269	-0.022685	-0.108521	0.011731	0.013224	-0.015400	-0.034831	-0.001505
Solids	-0.087615	-0.053269	1.000000	-0.051789	-0.162769	-0.005198	-0.005484	-0.015668	0.019409	0.040674
Chloramines	-0.024768	-0.022685	-0.051789	1.000000	0.006254	-0.028277	-0.023808	0.014990	0.013137	0.020784
Sulfate	0.010524	-0.108521	-0.162769	0.006254	1.000000	-0.016192	0.026776	-0.023347	-0.009934	-0.015303
Conductivity	0.014128	0.011731	-0.005198	-0.028277	-0.016192	1.000000	0.015647	0.004888	0.012495	-0.015496
Organic_carbon	0.028375	0.013224	-0.005484	-0.023808	0.026776	0.015647	1.000000	-0.005667	-0.015428	-0.015567
Trihalomethanes	0.018278	-0.015400	-0.015668	0.014990	-0.023347	0.004888	-0.005667	1.000000	-0.020497	0.009244
Turbidity	-0.035849	-0.034831	0.019409	0.013137	-0.009934	0.012495	-0.015428	-0.020497	1.000000	0.022682
Potability	0.014530	-0.001505	0.040674	0.020784	-0.015303	-0.015496	-0.015567	0.009244	0.022682	1.000000

[]:

Project Conclusion:

In conclusion, data analytics is an indispensable tool in the field of water quality analysis, enabling us to gain valuable insights into the health and safety of water sources. By harnessing the power of data analytics, we can make informed decisions that impact public health, environmental sustainability, and resource management







