College code: 4212

Register num: 421221243019

WATER QUALITY ANALYSIS

DATA ANALYTICS WITH COGNOS:GROUP2

PHASE:3

This phase involves in designing of the steps that defining in each phase of the previous documentation this involves importing necessary functions, data processing and so on in this phase we have to begin our project by loading and preprocessing the dataset.

The IBM suggests using the jupyter notebook for loading and preprocess the dataset:

Here for this project title we need to define the loading the libraries, understand the data and visualize the missing values.

For this certain inputs are defined for this project.in this phase each of the input lines of the project is given as follows:

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phase3

October 18, 2023

```
import pandas as pd
     import numpy as np
[2]: | df = pd.read_csv("water_potability.csv")
[3]: df.head
[3]: <bound method NDFrame.head of
                                                        Hardness
                                                                        Solids
                                                  ph
     Chloramines
                      Sulfate \
     0
                NaN
                      204.890456
                                   20791.31898
                                                    7.300212
                                                               368.516441
     1
                                                    6.635246
           3.716080
                      129.422921
                                   18630.05786
                                                                      NaN
     2
           8.099124
                      224.236259
                                   19909.54173
                                                    9.275884
                                                                      NaN
     3
           8.316766
                      214.373394
                                   22018.41744
                                                    8.059332
                                                               356.886136
           9.092223
                      181.101509
                                   17978.98634
                                                    6.546600
                                                               310.135738
                                   47580.99160
     3271
           4.668102
                      193.681736
                                                    7.166639
                                                               359.948574
     3272
           7.808856
                      193.553212
                                   17329.80216
                                                    8.061362
                                                                      NaN
     3273
           9.419510
                                                                      NaN
                      175.762646
                                   33155.57822
                                                    7.350233
     3274 5.126763
                      230.603758
                                   11983.86938
                                                    6.303357
                                                                      NaN
     3275 7.874671
                      195.102299
                                   17404.17706
                                                    7.509306
                                                                      NaN
           Conductivity
                          Organic_carbon
                                           Trihalomethanes
                                                              Turbidity
                                                                         Potability
     0
             564.308654
                                10.379783
                                                  86.990970
                                                               2.963135
                                                                                   0
     1
             592.885359
                                                                                   0
                                15.180013
                                                  56.329076
                                                               4.500656
     2
                                16.868637
             418.606213
                                                  66.420093
                                                               3.055934
                                                                                   0
     3
             363.266516
                                18.436525
                                                 100.341674
                                                               4.628771
             398.410813
                                11.558279
                                                  31.997993
                                                               4.075075
     3271
             526.424171
                                                  66.687695
                                                               4.435821
                                13.894419
                                                                                   1
     3272
             392.449580
                                19.903225
                                                        {\tt NaN}
                                                               2.798243
                                                                                   1
     3273
             432.044783
                                                  69.845400
                                                               3.298875
                                                                                   1
                                11.039070
     3274
             402.883113
                                11.168946
                                                  77.488213
                                                               4.708658
                                                                                   1
     3275
             327.459761
                                                  78.698446
                                                               2.309149
                                16.140368
     [3276 rows x 10 columns]>
[4]:
    df.info(memory_usage="deep")
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3276 entries, 0 to 3275
Data columns (total 10 columns):

#	Column	Non-Null Count	Dtype	
0	ph	2785 non-null	float64	
1	Hardness	3276 non-null	float64	
2	Solids	3276 non-null	float64	
3	Chloramines	3276 non-null	float64	
4	Sulfate	2495 non-null	float64	
5	Conductivity	3276 non-null	float64	
6	Organic_carbon	3276 non-null	float64	
7	Trihalomethanes	3114 non-null	float64	
8	Turbidity	3276 non-null	float64	
9	Potability	3276 non-null	int64	
1. (1) (1)				

dtypes: float64(9), int64(1)
memory usage: 256.1 KB

```
[5]: print(df.shape)
    print(len(df))
    print(f'Number of rows: {df.shape[0]} \nNumber of columns: {df.shape[1]}')
```

(3276, 10) 3276

Number of rows: 3276 Number of columns: 10

[6]: df.describe()

[6]:		ph	Hardness	Solids	Chloramines	Sulfate	\
	count	2785.000000	3276.000000	3276.000000	3276.000000	2495.000000	
	mean	7.080795	196.369496	22014.092526	7.122277	333.775777	
	std	1.594320	32.879761	8768.570828	1.583085	41.416840	
	min	0.000000	47.432000	320.942611	0.352000	129.000000	
	25%	6.093092	176.850538	15666.690300	6.127421	307.699498	
	50%	7.036752	196.967627	20927.833605	7.130299	333.073546	
	75%	8.062066	216.667456	27332.762125	8.114887	359.950170	
	max	14.000000	323.124000	61227.196010	13.127000	481.030642	

	Conductivity	Organic_carbon	Trihalomethanes	Turbidity	Potability
count	3276.000000	3276.000000	3114.000000	3276.000000	3276.000000
mean	426.205111	14.284970	66.396293	3.966786	0.390110
std	80.824064	3.308162	16.175008	0.780382	0.487849
min	181.483754	2.200000	0.738000	1.450000	0.000000
25%	365.734414	12.065801	55.844536	3.439711	0.000000
50%	421.884968	14.218338	66.622485	3.955028	0.000000
75%	481.792305	16.557652	77.337473	4.500320	1.000000

753.342620 28.300000 124.000000 6.739000 1.000000 max [7]: df.describe? [8]: df.isnull().sum() [8]: ph 491 Hardness 0 Solids 0 Chloramines 0 Sulfate 781 Conductivity 0 Organic carbon 0 Trihalomethanes 162 Turbidity 0 Potability 0 dtype: int64 [9]: def isnull_prop(df): total_rows = df.shape[0] missing_val_dict = {} for col in df.columns: missing_val_dict[col] = [df[col].isnull().sum(), (df[col].isnull(). ⇒sum() / total_rows)] return missing_val_dict null_dict = isnull_prop(df) print(null_dict.items()) dict_items([('ph', [491, 0.14987789987789987]), ('Hardness', [0, 0.0]), ('Solids', [0, 0.0]), ('Chloramines', [0, 0.0]), ('Sulfate', [781, 0.23840048840048841]), ('Conductivity', [0, 0.0]), ('Organic_carbon', [0, 0.0]), ('Trihalomethanes', [162, 0.04945054945054945]), ('Turbidity', [0, 0.0]), ('Potability', [0, 0.0])]) [10]: df_missing = pd.DataFrame.from_dict(null_dict, orient="index", columns=['missing', 'miss_percent']) df_missing

[10]:		missing	miss_percent
	ph	491	0.149878
	Hardness	0	0.000000
	Solids	0	0.000000
	Chloramines	0	0.000000
	Sulfate	781	0.238400
	Conductivity	0	0.000000
	Organic carbon	0	0.000000

Trihalomethanes	162	0.049451
Turbidity	0	0.000000
Potability	0	0.000000