PROJECT DEVELOPMENT PHASE - SPRINT II

Assignment Date	06 October 2022
Team ID	PNT2022TMID25561
Project Name	Efficient Water Quality Analysis and Prediction using Machine Learning
Maximum Marks	8 Mark

DATA PRE-PROCESSING

Click here to view the project:

Importing Required Package:

```
import pandas as pd
import seaborn as sns
import numpy as np
from matplotlib import pyplot as plt
%matplotlib inline
```

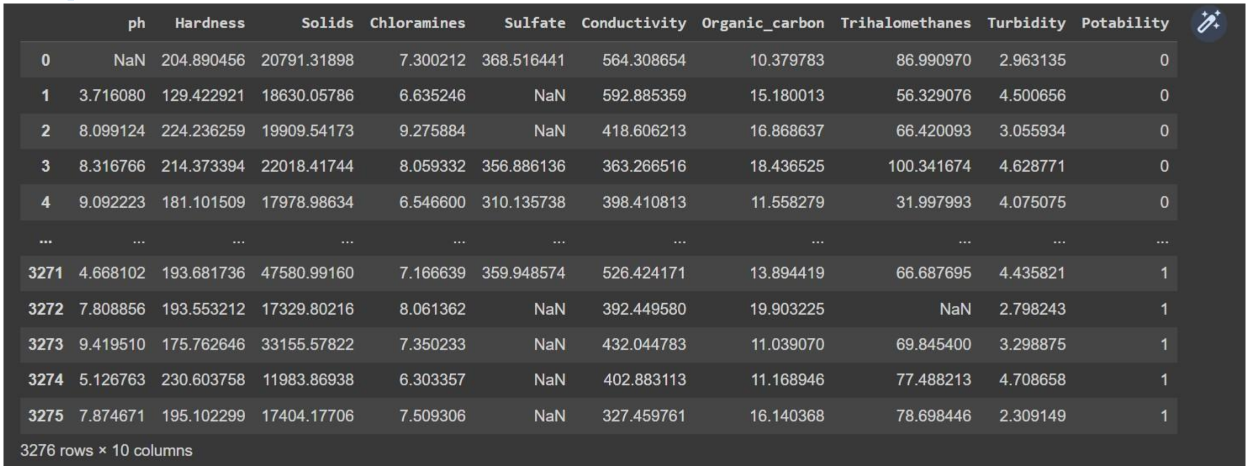
Loading the Dataset

SOLUTION:

```
df = pd.read_csv("water_potability.csv")

df
```

Output:



```
            ph    Hardness        Solids  Chloramines     Sulfate  Conductivity  Organic_carbon  Trihalomethanes  Turbidity  Potability
0          NaN  204.890456   20791.31898     7.300212  368.516441    564.308654       10.379783        86.990970   2.963135           0
1     3.716080  129.422921   18630.05786     6.635246         NaN    592.885359       15.180013        56.329076   4.500656           0
2     8.099124  224.236259   19909.54173     9.275884         NaN    418.606213       16.868637        66.420093   3.055934           0
3     8.316766  214.373394   22018.41744     8.059332  356.886136    363.266516       18.436525       100.341674   4.628771           0
4     9.092223  181.101509   17978.98634     6.546600  310.135738    398.410813       11.558279        31.997993   4.075075           0
...        ...         ...           ...          ...         ...           ...             ...              ...        ...         ...
3271  4.668102  193.681736   47580.99160     7.166639  359.948574    526.424171       13.894419        66.687695   4.435821           1
3272  7.808856  193.553212   17329.80216     8.061362         NaN    392.449580       19.903225              NaN   2.798243           1
3273  9.419510  175.762646   33155.57822     7.350233         NaN    432.044783       11.039070        69.845400   3.298875           1
3274  5.126763  230.603758   11983.86938     6.303357         NaN    402.883113       11.168946        77.488213   4.708658           1
3275  7.874671  195.102299   17404.17706     7.509306         NaN    327.459761       16.140368        78.698446   2.309149           1
3276 rows × 10 columns
```

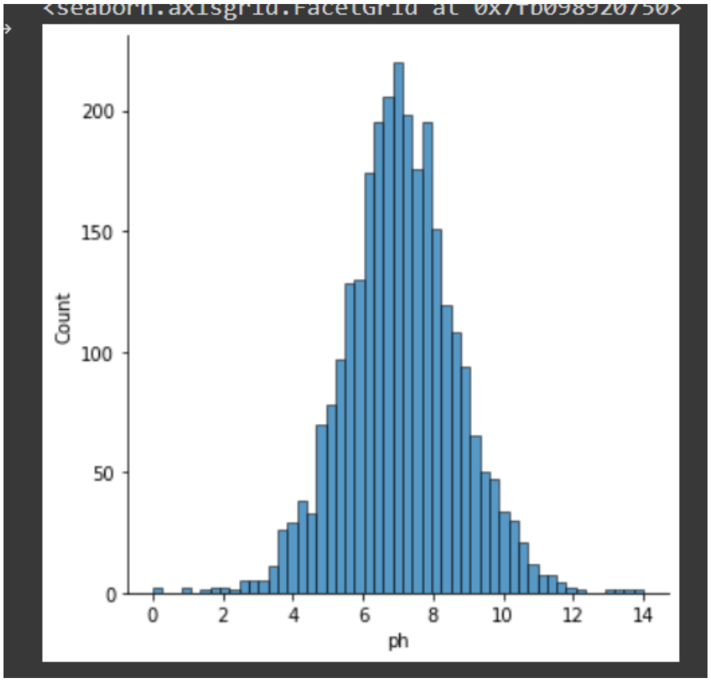
Visualizations

Univariate Analysis

Solution:

```
sns.displot(df.ph)
```

Output:



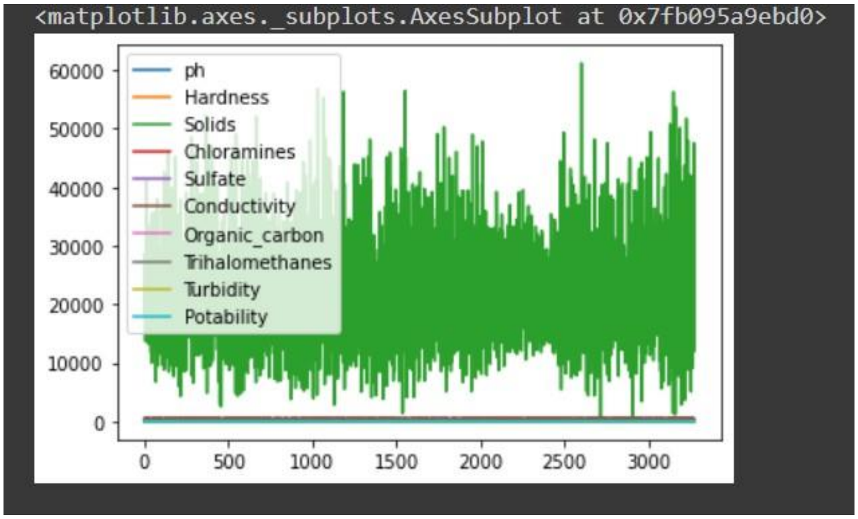
```
<seaborn.axisgrid.FacetGrid at 0x7fb098920750>
```

Bi-Variate Analysis

Solution:

```
df.plot.line()
```

Output:



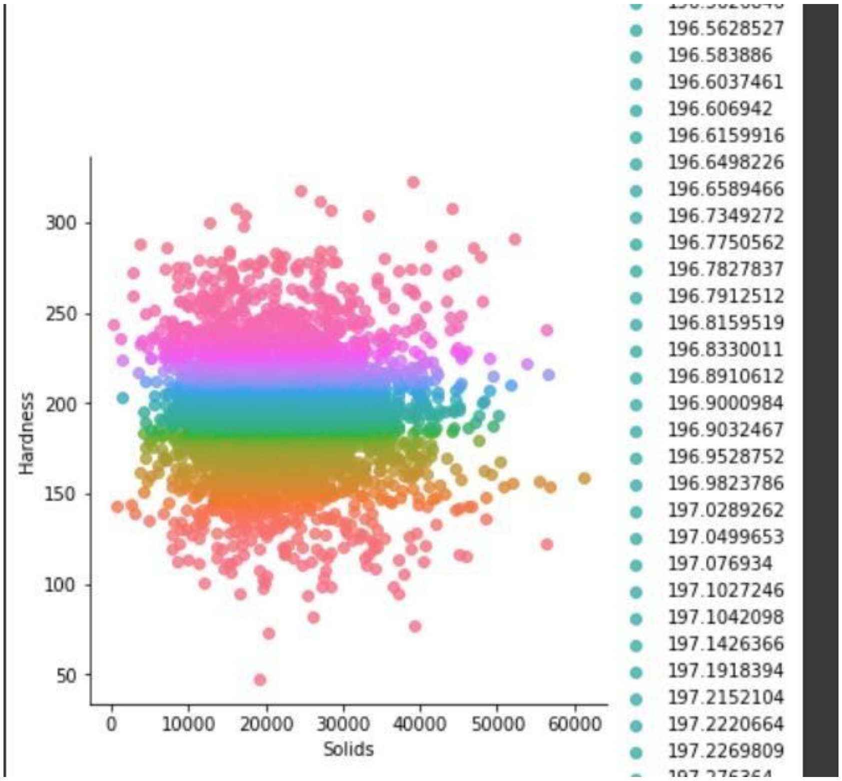
```
<matplotlib.axes._subplots.AxesSubplot at 0x7fb095a9ebd0>
```

Multi - Variate Analysis

Solution:

```
sns.lmplot(“Solids”,"Hardness",df,hue="Hardness", fit_reg=False);
```

Output:

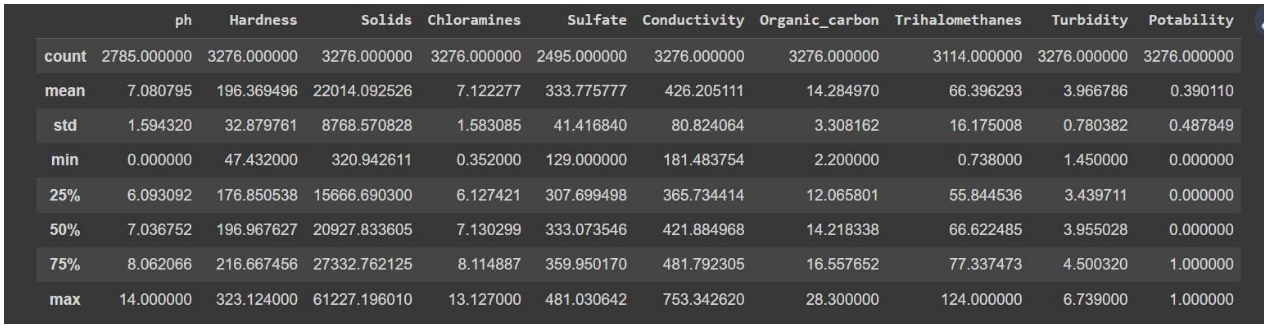


. Perform descriptive statistics on the dataset.

Solution:

```
df.describe()
```

Output:



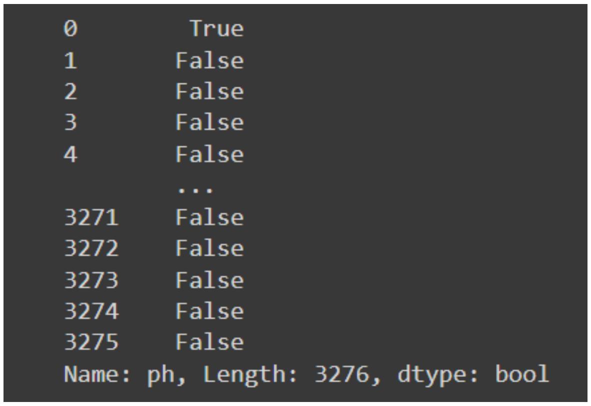
	ph	Hardness	Solids	Chloramines	Sulfate	Conductivity	Organic_carbon	Trihalomethanes	Turbidity	Potability
count	2785.000000	3276.000000	3276.000000	3276.000000	2495.000000	3276.000000	3276.000000	3114.000000	3276.000000	3276.000000
mean	7.080795	196.369496	22014.092526	7.122277	333.775777	426.205111	14.284970	66.396293	3.966786	0.390110
std	1.594320	32.879761	8768.570828	1.583085	41.416840	80.824064	3.308162	16.175008	0.780382	0.487849
min	0.000000	47.432000	320.942611	0.352000	129.000000	181.483754	2.200000	0.738000	1.450000	0.000000
25%	6.093092	176.850538	15666.690300	6.127421	307.699498	365.734414	12.065801	55.844536	3.439711	0.000000
50%	7.036752	196.967627	20927.833605	7.130299	333.073546	421.884968	14.218338	66.622485	3.955028	0.000000
75%	8.062066	216.667456	27332.762125	8.114887	359.950170	481.792305	16.557652	77.337473	4.500320	1.000000
max	14.000000	323.124000	61227.196010	13.127000	481.030642	753.342620	28.300000	124.000000	6.739000	1.000000

Handle the Missing values.

Solution:

```
data = pd.read_csv("water_potability.csv")
pd.isnull(data["ph"])
```

Output:



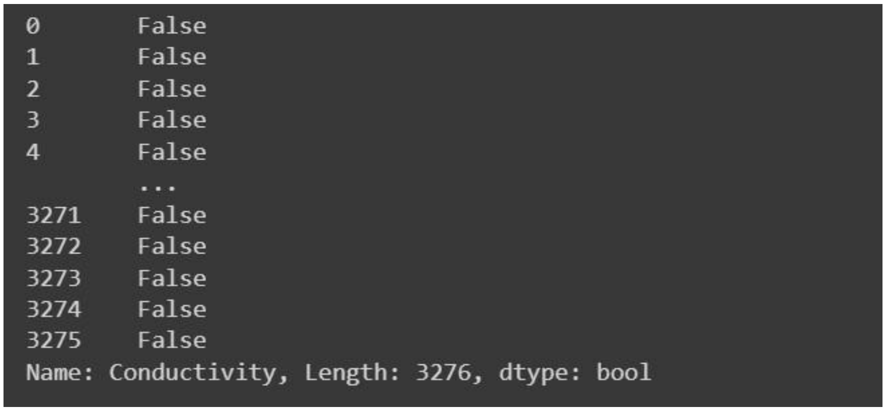
```
0        True
1       False
2       False
3       False
4       False
        ...
3271    False
3272    False
3273    False
3274    False
3275    False
Name: ph, Length: 3276, dtype: bool
```

Handling Missing Values -2

Solution:

```
data = pd.read_csv("water_potability.csv")
pd.isnull(data["conductivity"])
```

Output:



```
0       False
1       False
2       False
3       False
4       False
        ...
3271    False
3272    False
3273    False
3274    False
3275    False
Name: Conductivity, Length: 3276, dtype: bool
```

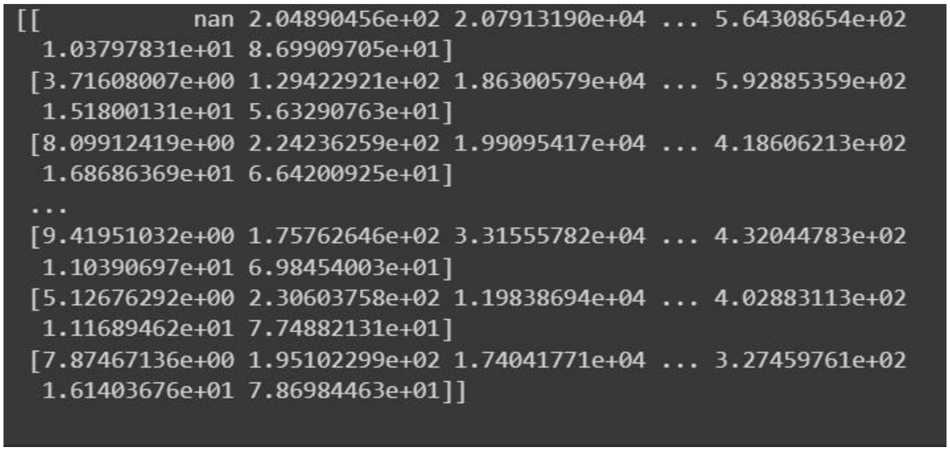
Split the data into dependent and independent variables

Split the data into Independent variables.

Solution:

```
X = df.iloc[:, :-2].values
print(X)
```

Output:



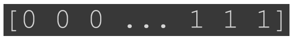
```
[[           nan 2.04890456e+02 2.07913190e+04 ... 5.64308654e+02
  1.03797831e+01 8.69909705e+01]
 [3.71608007e+00 1.29422921e+02 1.86300579e+04 ... 5.92885359e+02
  1.51800131e+01 5.63290763e+01]
 [8.09912419e+00 2.24236259e+02 1.99095417e+04 ... 4.18606213e+02
  1.68686369e+01 6.64200925e+01]
 ...
 [9.41951032e+00 1.75762646e+02 3.31555782e+04 ... 4.32044783e+02
  1.10390697e+01 6.98454003e+01]
 [5.12676292e+00 2.30603758e+02 1.19838694e+04 ... 4.02883113e+02
  1.11689462e+01 7.74882131e+01]
 [7.87467136e+00 1.95102299e+02 1.74041771e+04 ... 3.27459761e+02
  1.61403676e+01 7.86984463e+01]]
```

Split the data into Dependent variables.

Solution:

```
Y = df.iloc[:, -1].values
print(Y)
```

Output:



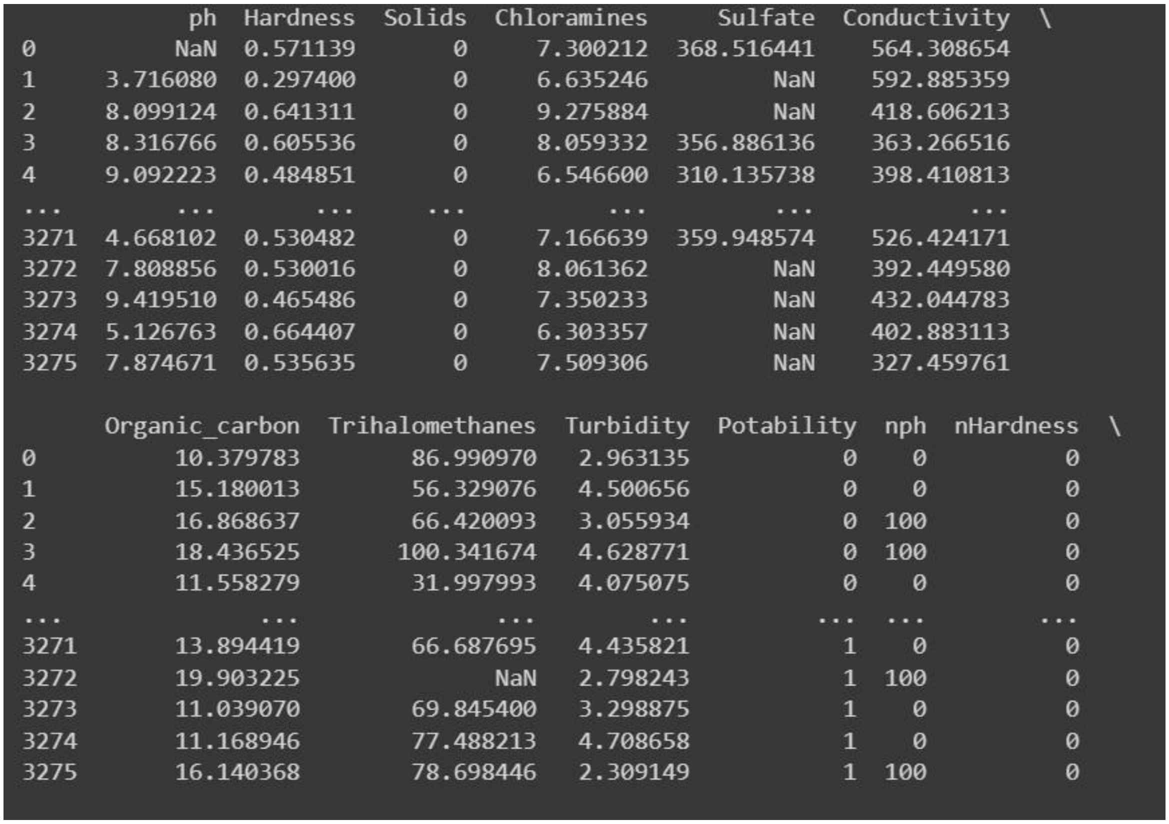
```
[0 0 0 ... 1 1 1]
```

Scale the independent variables

Solution:

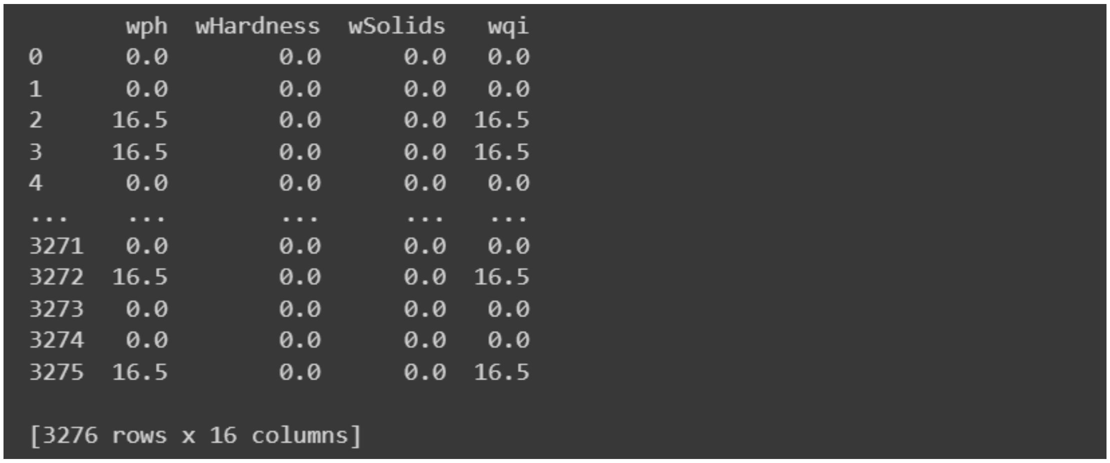
```
import pandas as pd
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
df[["Hardness"]] = scaler.fit_transform(df[["Hardness"]])
print(df)
```

Output:



```
            ph  Hardness  Solids  Chloramines     Sulfate  Conductivity  \
0          NaN  0.571139       0     7.300212  368.516441    564.308654
1     3.716080  0.297400       0     6.635246         NaN    592.885359
2     8.099124  0.641311       0     9.275884         NaN    418.606213
3     8.316766  0.605536       0     8.059332  356.886136    363.266516
4     9.092223  0.484851       0     6.546600  310.135738    398.410813
...        ...       ...     ...          ...         ...           ...
3271  4.668102  0.530482       0     7.166639  359.948574    526.424171
3272  7.808856  0.530016       0     8.061362         NaN    392.449580
3273  9.419510  0.465486       0     7.350233         NaN    432.044783
3274  5.126763  0.664407       0     6.303357         NaN    402.883113
3275  7.874671  0.535635       0     7.509306         NaN    327.459761

      Organic_carbon  Trihalomethanes  Turbidity  Potability  nph  nHardness  \
0          10.379783        86.990970   2.963135           0    0          0
1          15.180013        56.329076   4.500656           0    0          0
2          16.868637        66.420093   3.055934           0  100          0
3          18.436525       100.341674   4.628771           0  100          0
4          11.558279        31.997993   4.075075           0    0          0
...              ...              ...        ...         ...  ...        ...
3271       13.894419        66.687695   4.435821           1    0          0
3272       19.903225              NaN   2.798243           1  100          0
3273       11.039070        69.845400   3.298875           1    0          0
3274       11.168946        77.488213   4.708658           1    0          0
3275       16.140368        78.698446   2.309149           1  100          0
```



```
       wph  wHardness  wSolids   wqi
0      0.0        0.0      0.0   0.0
1      0.0        0.0      0.0   0.0
2     16.5        0.0      0.0  16.5
3     16.5        0.0      0.0  16.5
4      0.0        0.0      0.0   0.0
...    ...        ...      ...   ...
3271   0.0        0.0      0.0   0.0
3272  16.5        0.0      0.0  16.5
3273   0.0        0.0      0.0   0.0
3274   0.0        0.0      0.0   0.0
3275  16.5        0.0      0.0  16.5

[3276 rows x 16 columns]
```

Split the data into training and testing

Solution:

```
from sklearn.model_selection import train_test_split
train_size=0.8
X = df.drop(columns = ['ph']).copy()
y = df['ph']
```

```
X_train, X_rem, y_train, y_rem = train_test_split(X,y, train_size=0.8)
test_size = 0.5
X_valid, X_test, y_valid, y_test = train_test_split(X_rem,y_rem, test_size=0.
5)
print(X_train.shape), print(y_train.shape)
print(X_valid.shape), print(y_valid.shape)
print(X_test.shape), print(y_test.shape)
```

Output:

```
(2620, 9)
(2620,)
(328, 9)
(328,)
(328, 9)
(328,)
(None, None)
```

Water Quality Index Calculation :

Solution:

```
df['nph']=df.ph.apply(lambda x: (100 if (8.5>=x>=7)
else(80 if (8.6>=x>=8.5) or (6.9>=x>=6.8)
else(60 if (8.8>=x>=8.6) or (6.8>=x>=6.7)
else(40 if (9>=x>=8.8) or (6.7>=x>=6.5)
else 0)))))
```

For second column:

```
df['nHardness']=df.Hardness.apply(lambda x: (100 if (x>=6)
else(80 if (6>=x>=5.1)

else(60 if (5>=x>=4.1)

else(40 if (4>=x>=3)

else 0)))))
```

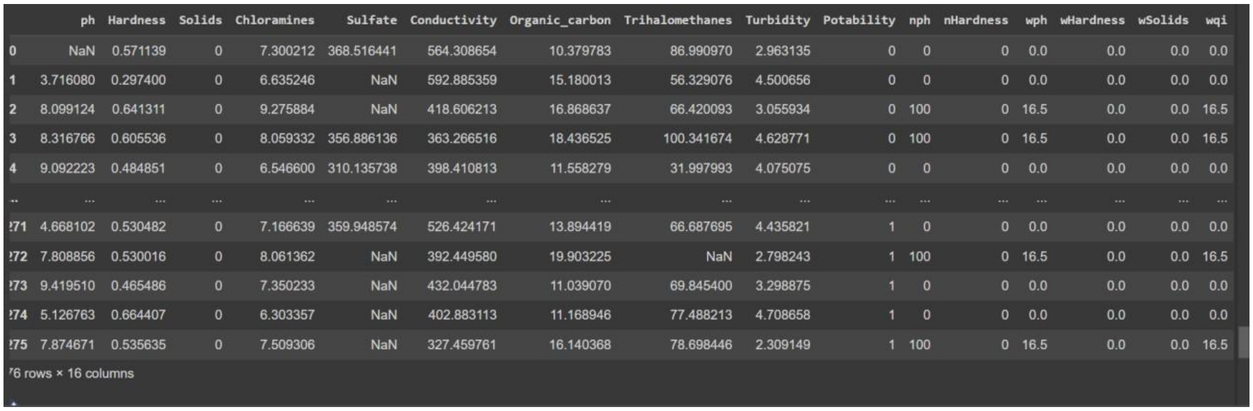
For Third Column:

```
df['Solids']=df.Solids.apply(lambda x:(100 if (5>=x>=0)
else(80 if (50>=x>=5)
else(60 if (500>=x>=50)
else(40 if (10000>=x>=500)
else 0)))))
```

Calculation water Quality Index:

```
#calculation of water quality index WQI
df['wph']=df.nph*0.165
df['wHardness']=df.nHardness*0.281
df['wSolids']=df.Solids*0.281
df['wqi']=df.wph+df.wHardness+df.wSolids
df
```

Output:



	ph	Hardness	Solids	Chloramines	Sulfate	Conductivity	Organic_carbon	Trihalomethanes	Turbidity	Potability	nph	nHardness	wph	wHardness	wSolids	wqi
0	NaN	0.571139	0	7.300212	368.516441	564.308654	10.379783	86.990970	2.963135	0	0	0	0.0	0.0	0.0	0.0
1	3.716080	0.297400	0	6.635246	NaN	592.885359	15.180013	56.329076	4.500656	0	0	0	0.0	0.0	0.0	0.0
2	8.099124	0.641311	0	9.275884	NaN	418.606213	16.868637	66.420093	3.055934	0	100	0	16.5	0.0	0.0	16.5
3	8.316766	0.605536	0	8.059332	356.886136	363.266516	18.436525	100.341674	4.628771	0	100	0	16.5	0.0	0.0	16.5
4	9.092223	0.484851	0	6.546600	310.135738	398.410813	11.558279	31.997993	4.075075	0	0	0	0.0	0.0	0.0	0.0
..	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
!71	4.668102	0.530482	0	7.166639	359.948574	526.424171	13.894419	66.687695	4.435821	1	0	0	0.0	0.0	0.0	0.0
!72	7.808856	0.530016	0	8.061362	NaN	392.449580	19.903225	NaN	2.798243	1	100	0	16.5	0.0	0.0	16.5
!73	9.419510	0.465486	0	7.350233	NaN	432.044783	11.039070	69.845400	3.298875	1	0	0	0.0	0.0	0.0	0.0
!74	5.126763	0.664407	0	6.303357	NaN	402.883113	11.168946	77.488213	4.708658	1	0	0	0.0	0.0	0.0	0.0
!75	7.874671	0.535635	0	7.509306	NaN	327.459761	16.140368	78.698446	2.309149	1	100	0	16.5	0.0	0.0	16.5

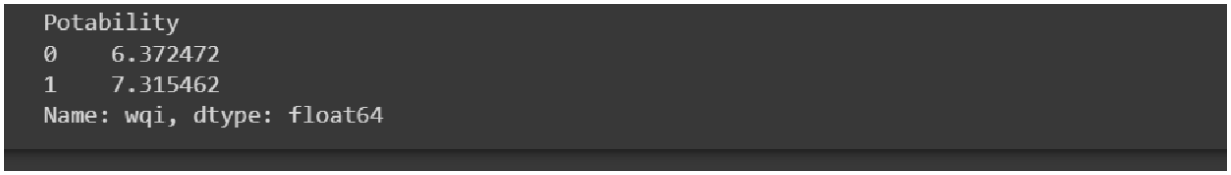
'6 rows × 16 columns

Calculate the Average of WQI:

Solution:

```
average=df.groupby('Potability')['wqi'].mean()
```

Output:



```
Potability
0    6.372472
1    7.315462
Name: wqi, dtype: float64
```