Water Quality Prediction

CSP 571 Final Report

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Abstract

Water quality is a significant concern in India, with numerous states facing challenges in managing and maintaining clean water sources. To tackle this issue, machine learning algorithms are being employed to cluster states based on their water quality parameters. This study aims to cluster Indian states based on water quality data using machine learning algorithms. The dataset includes several water quality parameters such as pH, total dissolved solids, and hardness, among others. The data was preprocessed and scaled, and then k-means clustering was applied to the dataset. We have divided states into six clusters based on their water quality and air quality parameters. This study provides valuable insights into water quality in India and can aid policymakers in developing strategies to improve water quality in states with poor water quality.

1. Overview

Water and air pollution are significant challenges in India, affecting the health and well-being of its citizens. According to the Central Pollution Control Board (CPCB), more than half of India's groundwater is contaminated with pollutants such as arsenic, fluoride, and iron, with states like Bihar, Uttar Pradesh, West Bengal, and Rajasthan having the worst water quality. India is also known for having some of the most polluted air in the world, with high levels of particulate matter, nitrogen dioxide, and sulfur dioxide. States like Delhi, Uttar Pradesh, Bihar, and West Bengal are among the worst affected.

The sources of water and air pollution in India are diverse and include industrial effluents, domestic sewage, agricultural runoff, and solid waste disposal for water pollution and vehicular emissions, industrial emissions, construction activities, and biomass burning for air pollution. Exposure to polluted water and air can cause severe health impacts, including respiratory diseases, cancer, neurological disorders, and developmental problems.

To address these issues, the Indian government has launched various initiatives, including the National Clean Air Programme (NCAP) and the Jal Jeevan Mission (JJM), to improve water and air quality. However, infrastructure inadequacy, inadequate enforcement of regulations, and a growing population continue to pose significant challenges to improving water and air quality in India.

2. Objectives:

- Early detection of water contamination: The model can be trained to detect contaminants in water at an early stage, which can help prevent widespread contamination and protect public health
- Water quality monitoring: This model can be used to monitor water quality in real-time, allowing water treatment plants to take corrective action quickly if the water quality falls below acceptable levels
- **Air quality monitoring:** This model can be used to monitor the air quality and provides information on the level of air pollution across different states in India

3. Data Collection & Processing

Data Collection: We collected water quality data that includes the values of environmental parameters (temperature, dissolved oxygen, pH, conductivity, BOD, Nitrate N + Nitrite N, Fecal Coliform, Total Coliform) from http://www.cpcbenvis.nic.in/water_quality_data.html

- To Support the features of water quality we have extracted air quality data also to see if it adds some purpose while clustering from http://www.cpcbenvis.nic.in/air_quality_data.html
- In total, we have **Water and Air Quality data** for the years 2017 to 2021 in the form of pdf files for each year
- We collated all the pdfs and combined them into .xlsx files so that they can be processed in R.

Data Processing:

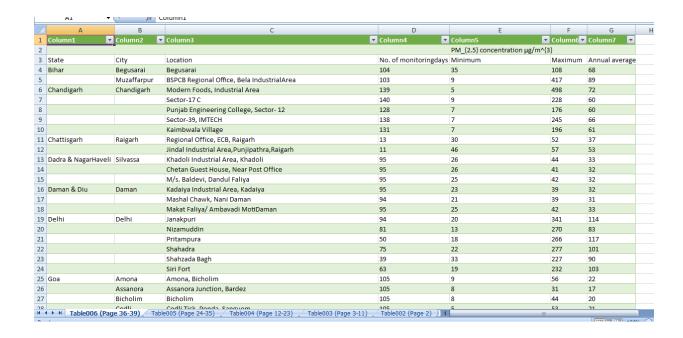
Snapshot of Air Quality Data:

SAVALI AT THOTAPALLI REGULATOR, VIZIANAGARAM WSADHARA, KALINGAPATNAM, VIZIANAGARAM	State Name ANDHRA PRADESH	Min	Max	Min	Max	0.00
USADHARA, KALINGAPATNAM, VIZIANAGARAM		20.0			IVIAA	Min
		26.0	31.0	6.0	8.0	6.6
	ANDHRA PRADESH	26.0	29.0	6.2	8.2	6.2
MANGANGA AT D/S OF MADHUBAN,DAMAN	DAMAN AND DIU, DADRAAND NAGAR HAVELI	27.0	30.0	6.0	16.9	7.8
MANGANGA AT ZARI CAUSE WAYBRIDGE, DAMAN	DAMAN AND DIU, DADRAAND NAGAR HAVELI	26.0	30.0	4.2	6.6	7.3
MANGANGA AT DISCHARGE POINT OFDISTILLERY, DAMAN	DAMAN AND DIU, DADRAAND NAGAR HAVELI	26.0	30.0	4.3	6.2	7.3
MANGANGA AT DAMAN JETTY, MOTIDAMAN	DAMAN AND DIU, DADRAAND NAGAR HAVELI	26.0	30.0	5.2	6.8	7.2
MANGANGA AT VAPI WEIR, VAPI,DAMAN	DAMAN AND DIU, DADRAAND NAGAR HAVELI	27.0	30.0	4.6	12.8	7.4
MANGANGA AT LAVACHA TEMPLE,SILVASSA	DAMAN AND DIU, DADRAAND NAGAR HAVELI	25.0	30.0	5.2	6.9	7.9
MANGANGA AT D/S OF M/S SURATBEVERAGES, VILLAGE DADRA, SILVASSA	DAMAN AND DIU, DADRAAND NAGAR HAVELI	27.0	30.0	4.5	6.6	7.9
MANGANGA AT NAROLI BRIDGE, SILVASSA	DAMAN AND DIU, DADRAAND NAGAR HAVELI	25.0	30.0	5.6	7.3	7.9
MANGANGA AT VILLAGE NAMDHA, VAPI	DAMAN AND DIU, DADRAAND NAGAR HAVELI	26.0	30.0	4.4	8.6	7.4
RI AT D/S OF PT. WHEREKUMBARJRIA CANAL JOINS, GOA	GOA	27.0	34.0	4.9	7.6	6.8
NDOVI AT NEGHBOURHOOD OFPANAJI, GOA	GOA	26.0	32.0	4.5	7.8	6.9
ARI AT PANCHAWADI	GOA	29.0	34.0	4.2	7.5	6.0
NDOVI AT TONCA, MARCELA, GOA	GOA	25.0	34.0	4.3	7.8	5.9
ER KALNA AT CHANDEL- PERNEM,GOA	GOA	26.0	30.0	6.4	7.7	6.1
ER VALVANT AT SANKLI - BICHOLIM,GOA	GOA	27.0	32.0	6.3	8.3	6.0
ER MADAI AT DABOS - VALPOI, GOA	GOA	27.0	31.5	6.0	8.4	6.0
ER KHANDEPAR AT OPA - PONDA,GOA	GOA	27.6	34.0	6.8	7.6	6.1
ER TALPONA AT CANACONA, GOA	GOA	27.0	34.0	4.9	8.1	5.8
ER ASSONORA AT ASSONORA, GOA	GOA	27.9	33.0	5.7	7.4	5.8
ER KHANDEPAR AT CODLI NEARBRIDGE , U/S OPA WATERWORKS, SANGUEM	GOA	28.0	34.0	6.7	7.7	6.1
ER SAL PAZORKHONI,CUNCOLIM(NEAR CULVERT MARGAO-CANACONA NATIONAL HIGHW	GOA	28.0	32.0	4.9	7.0	5.9
ER KUSHAWATI NEAR BUND ATKEVONA, RIVON, SANGUEM	GOA	25.0	34.0	7.0	7.8	6.0
CD CAL MEAD HOTEL LEELA MODOD CAVELOCCIM	COA	24.2	24.0	20	0.6	71
MANA AR NI EFFEFFEFFEFFEFFEFFEFFFFFFFFFFFFFFFFF	ANGANGA AT ZARI CAUSE WAYBRIDGE, DAMAN ANGANGA AT DISCHARGE POINT OF DISTILLERY, DAMAN ANGANGA AT DAMAN JETTY, MOTIDAMAN ANGANGA AT DAMAN JETTY, MOTIDAMAN ANGANGA AT VAPI WEIR, VAPI,DAMAN ANGANGA AT VAPI WEIR, VAPI,DAMAN ANGANGA AT LAVACHA TEMPLE,SILVASSA ANGANGA AT NAROLI BRIDGE,SILVASSA ANGANGA AT NILLAGE NAMDHA,VAPI RI AT D/S OF PT. 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Data cleaning: The data was in different formats for each year. We converted and processed the data in the same format so it's easy to process this file.

Snapshot of Air Quality Data:

Most of our efforts in the project went into processing this file because its not in a proper format with multiple sheets in excel. More over each sheet have particular feature for example in sheet1 PM2.5 values are there and in sheet2 PM10 values are there.



Moreover, if we see there are blank values in state and city because for example Bihar state is repeated twice in 4 and 5 rows so, the 5 row has state value as blank. We need to fill this and for each record we are flagging it's measure which is present in column 5 and pivoting it via mean of measures. Once this part is done we are taking means and filling blank values in our data frame.

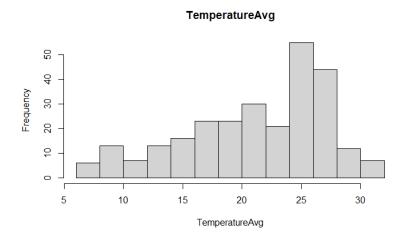
We have merged Water Quality of Data with Air Quality data based on state, city and year.

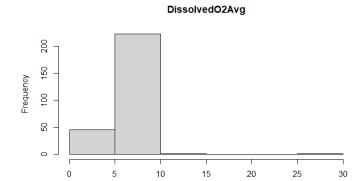
4. Data Analysis

We will present Analysis for 2017 Data and we have done the same analysis for rest of the years. We have got Water quality features like minimum and maximum values, we have averaged them and used while using clustering.

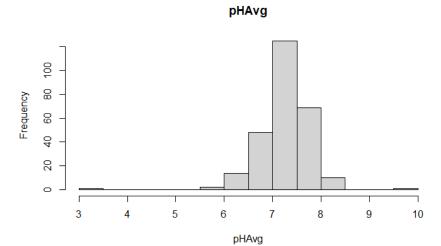
Exploratory Data Analysis (EDA):

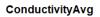
Visualizing the data for all features in the dataset including Water and Air Quality Features.

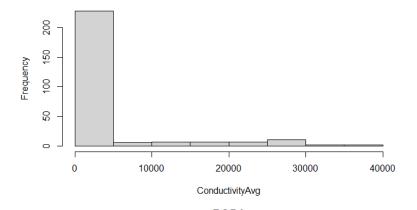


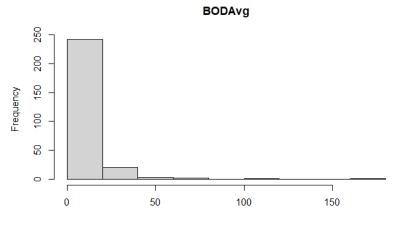


DissolvedO2Avg

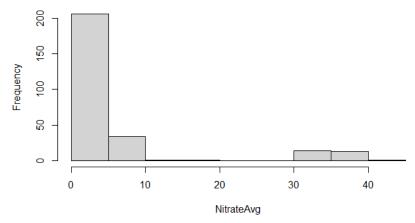




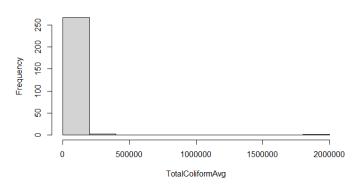




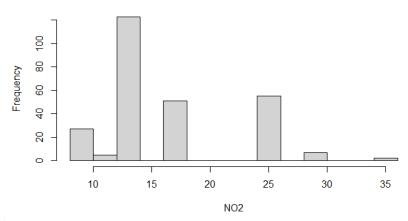


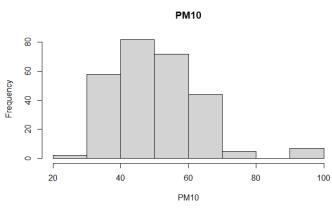


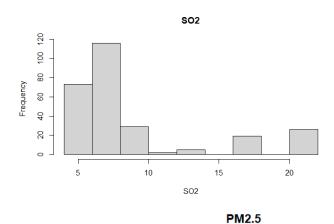
TotalColiformAvg

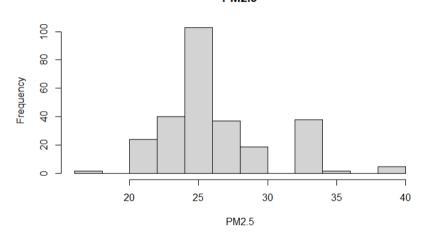


NO2



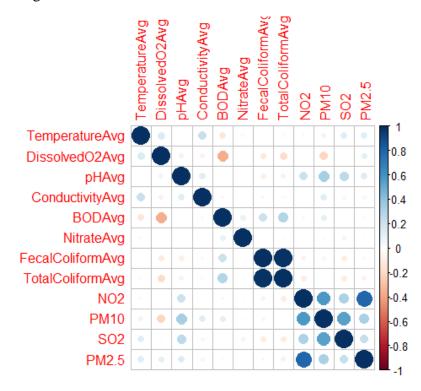






We can observe that most of the attributes are normally distributed, and some attributes are skewed. Anyways, we will perform scaling before using clustering models.

Obtaining the covariance matrix:



We can see that Total Coliform Avg and Fecal Coliform Avg are same and PM2.5, NO2 are highly

correlated. So, we will remove those columns for further processing.

Performing PCA for 2017 Data:

```
PC1
                                    PC2
                                               PC3
                                                           PC4
                                                                       PC5
                                                                                   PC6
                                                                                              PC7
                 0.35247931 -0.16609907 -0.41145789 0.12002435 -0.39078002 -0.28510478 0.22205628
TemperatureAvg
                 0.52797169 0.03667973 0.02672452 0.03326143
                                                                0.19101186 -0.14498752 -0.09851537
DissolvedO2Ava
                -0.03466755 -0.37731106 -0.22467152 -0.37981335
                                                                0.54072343
                                                                           0.23271310 -0.03703716
pHAVa
ConductivityAvg 0.13785851 -0.14080510 -0.62855619 0.38041477
                                                                0.02841395
                                                                           0.45370455
                                                                                       0.11568707
                0.36561378
BODAVQ
NitrateAva
                -0.25471868
                             0.22631481 0.07635337
                                                   0.68224764 0.12244317
                                                                           0.02010386 -0.22942017
TotalColiformAvg -0.40027504 0.18624164 -0.45735422 -0.02741876 -0.09246997 -0.27896640 -0.25133608
                -0.15916206 -0.42680513
                                         0.28226348
                                                    0.33024486
                                                                0.05635247 -0.16830899 0.62128963
NO2
PM10
                -0.32606103 -0.41622281 0.04576228 0.13507552
                                                                0.03491503 0.24828216 -0.21068987
                -0.02274704 -0.40321836 0.15283930 -0.11095656 -0.65885009 0.20479694 -0.32719662
502
                 0.02328897 \;\; -0.44967807 \;\; -0.08671307 \;\; 0.11879117 \;\; 0.21404475 \;\; -0.62533654 \;\; -0.37323578
PM2.5
                         PC8
                                    PC9
                                                 PC10
                                                             PC11
                 0.330605635 -0.47779751 -0.1346027970 -0.15954557
TemperatureAvg
Dissolved02Avg
                 0.182696546 0.05975104 0.7038177741
                                                       0.34432559
pHAVg
                 0.509216387 -0.01583569 -0.2109544778
                                                       0.09867213
ConductivityAvg -0.273985211
                                         0.0665735502 -0.02705993
                              0.34799400
                                         0.5024532930 -0.33824420
BODAVQ
                 0.197221634
                              0.24528958
NitrateAvq
                 0.566475552 0.05955844
                                         0.0002618284 -0.14190274
TotalColiformAvg -0.057626241 -0.09084282 -0.0567735701
                                                       0.65637647
                 0.003154439 0.09393151 -0.0604248614
NO<sub>2</sub>
                                                       0.41687158
PM10
                -0.207043318 -0.60938541
                                         0.4149958309 -0.08006764
502
                 0.266337041 0.35191481 0.0417708942 0.14703468
PM2.5
                -0.214542884 0.26609098 -0.0654879302 -0.27987343
 [1] 0.294633176 0.256715245 0.120933762 0.103176527 0.080930646 0.065953880 0.033703160 0.021155724 0.014902656
[10] 0.004966155 0.002929070
```

The first 4 principal components explain 77% of the Variance in Data.

5. Clustering

For water and air quality analysis in Indian states, we used use **k-means clustering** to group states based on their water quality parameters.

Step 1: We started by selecting the relevant water and air quality parameters, such as pH, total dissolved solids, and concentration of various contaminants.

Step 2: We then used k-means clustering for grouping states based on air & water parameters in datsets

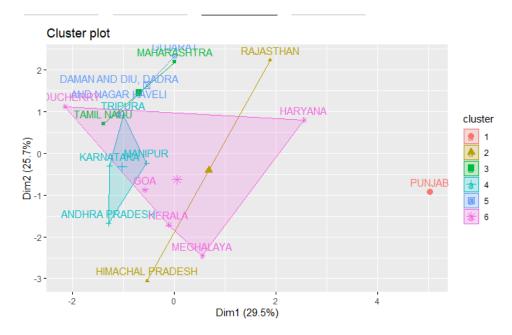
Step 3: We got with 6 clusters on our filtered dataset using K means clustering

Step 4: Then once the clusters were formed, we assigned ranks to the clusters based on their features like Fecal Coliform, BOD, Dissolved O2, PM10, PM2.5 and NO2 to give a meaning to these clusters. We took a reference of below criteria that is ideal for checking water quality:

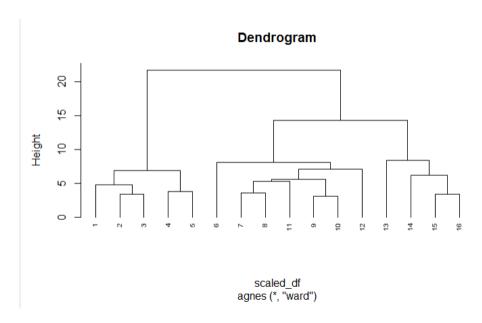
Designated-Best-Use	Class of water	Criteria	
Drinking WaterSource	A	Total Coliforms Organism MPN/100ml shall be 50 or less	
without conventional treatment but after disinfection		pH between 6.5 and 8.5	
		Dissolved Oxygen 6mg/l or more	
		Biochemical Oxygen Demand 5 days 20C 2mg/l or less	
Outdoor bathing (Organised)	В	Total Coliforms Organism MPN/100ml shall be 500 or less pH between 6.5 and 8.5 Dissolved Oxygen 5mg/l or more	
		Biochemical Oxygen Demand 5 days 20C 3mg/l or less	
Drinking water source after conventional treatment and	С	Total Coliforms Organism MPN/100ml shall be 5000 or less pH between 6 to 9 Dissolved Oxygen 4mg/l or more	
disinfection		Biochemical Oxygen Demand 5 days 20C 3mg/l or less	
Propagation of Wild life and	D	pH between 6.5 to 8.5 Dissolved Oxygen 4mg/l or more	
Fisheries		Free Ammonia (as N) 1.2 mg/l or less	
Irrigation, Industrial Cooling, Controlled Waste disposal	E	pH betwwn 6.0 to 8.5	
		Electrical Conductivity at 25C micro mhos/cm Max.2250	
Water not suitable for any purpose	F	Total Coliforms Organism greater than 5000 and pH is high	

K-means clustering can provide valuable insights into the water quality status of Indian states, such as identifying states with similar water quality profiles or identifying states with poor water quality that require immediate attention. However, like any clustering technique, the interpretation of results should be done with caution, and additional domain knowledge and contextual information should be taken into account. We have performed K Means clustering with 6 clusters on our filtered dataset. Then once the cluster are formed bas we have assigned ranks to the clusters based on their features like Fecal Coliform, BOD, Dissolved O2, PM10, PM2.5 and NO2.

We have obtained the following clusters:



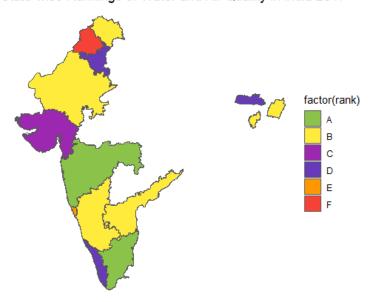
We have performed hierarchical clustering using ward's distance and we obtained same results with 6 clusters.



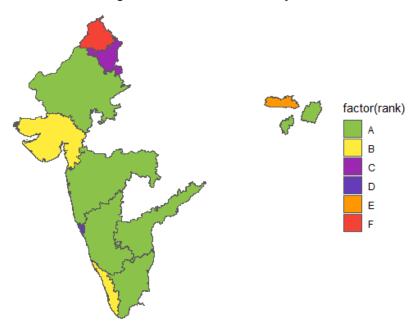
Year On Year Analysis:

We have did the same procedure for 5 years of data and obtained the following results via clustering.

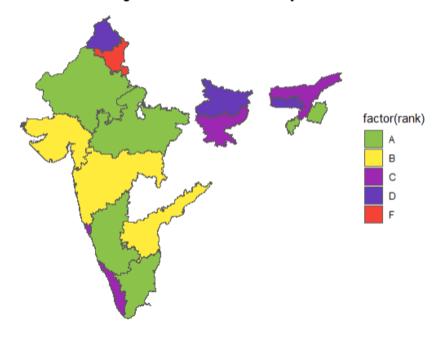
State-wise Rankings of Water and Air Quality in India 2017



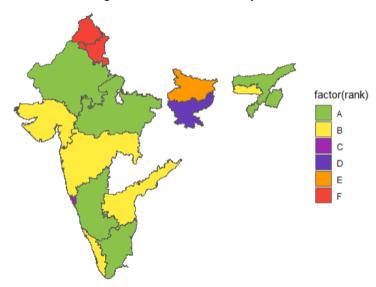
State-wise Rankings of Water and Air Quality in India 2018



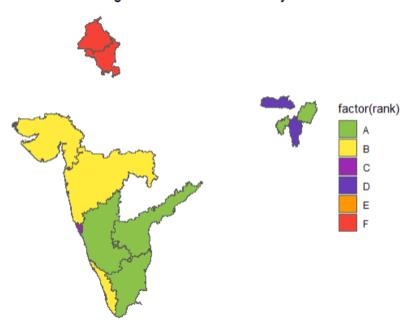
State-wise Rankings of Water and Air Quality in India 2019



State-wise Rankings of Water and Air Quality in India 2020



State-wise Rankings of Water and Air Quality in India 2021



Conclusion:

We had data from 16 states available. Further, we have clustered states based on their water and air quality. Two key observations:

- We can see that from 2017 most of the states are performing better in terms of water and air quality.
- We observed that Chandigarh and Haryana are performing very poor in terms of water and air quality and some action should be taken by government authorities to clean up the water

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