Problem Definition Description:

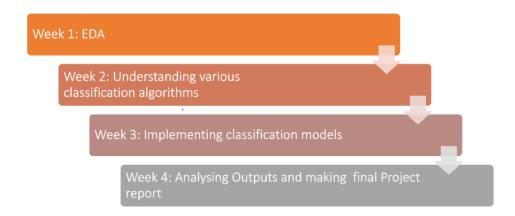
The presence of contaminants such as sewage waste, heavy metal, fertilizers, pesticides can be the cause of many illnesses in humans such as gastrointestinal illness, reproductive problems, and neurological disorders. Individuals with weakened immune systems may be at greater risk. Water testing is important for monitoring the safety of drinking water, investigation of disease outbreaks, validation process and preventive measures, especially in a country like India where 40% of the population is not expected to have clean water by 2030

The attempt here is to classify whether water is potable for drinking based on pH value, hardness of water, TDS (Total dissolved solids)value, Chloramines and sulphates content, conductivity, Total Organic Carbon (TOC), Trihalomethanes and turbidity of water and further to perform intense EDA and compare various classification models based on their prediction accuracy and effectiveness. This aims at the tapping of the cumulative effect of contaminants on the potability of water which is often overlooked

Technology landscape assessment:

In some regions, it has been shown that investments in water supply and sanitation can yield a net economic benefit, since the reductions in adverse health effects and health care costs outweigh the costs of undertaking the interventions. The general procedure for water quality assessment pivots around the investigative analysis followed by comparison to defined levels of each parameter, which neglects the obvious cumulative effect of chemicals/contaminants. Replacing Manual interventions with the deployed model could reduce overall cost of the project.

Project Planning:

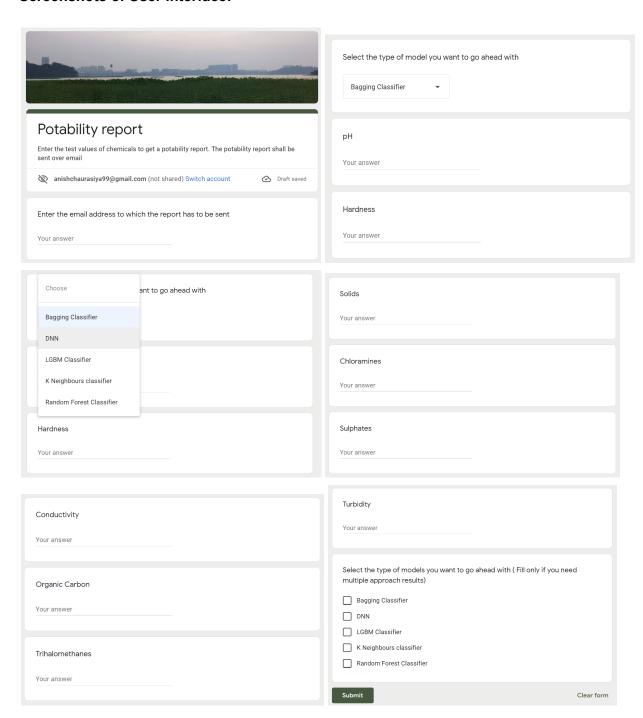


Conceptual Design

Processing End Existing Data Comparison ML classificaion from Available model for and choosing potability right model sources Data collection Water Data Input Data at multiple Quality to Model Preparation sources/hotspots Report

User End

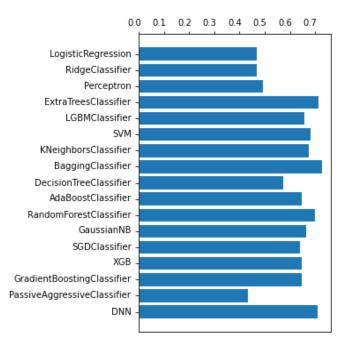
Screenshots of User interface:



Unit testing report:

0	LogisticRegression	0		
1	RidgeClassifier	0		
2	Perceptron	0		
3	ExtraTreesClassifier	0		
4	LGBMClassifier	0		
5	SVM	0		
6	KNeighborsClassifier	0		
7	BaggingClassifier	0		
8	DecisionTreeClassifier			
9	AdaBoostClassifier	0		
10	RandomForestClassifier	0		
11	GaussianNB	0		
12	SGDClassifier	0		
13	XGB	0		

Model training and testing report:



	Model Name	Precission	Recall	Accuracy
0	LogisticRegression	0.284211	0.409091	0.465
1	RidgeClassifier	0.284211	0.409091	0.465
2	Perceptron	0.336364	0.560606	0.490
3	ExtraTreesClassifier	0.586957	0.409091	0.710
4	LGBMClassifier	0.480519	0.560606	0.655
5	SVM	0.526316	0.303030	0.680
6	KNeighborsClassifier	0.530303	0.500000	0.670
7	BaggingClassifier	0.622222	0.424242	0.725
8	DecisionTreeClassifier	0.383721	0.500000	0.570
9	AdaBoostClassifier	0.454545	0.378788	0.645
10	RandomForestClassifier	0.543860	0.469697	0.695
11	GaussianNB	0.475000	0.287879	0.660
12	SGDClassifier	0.436364	0.363636	0.635
13	XGB	0.333333	0.075758	0.645
14	GradientBoostingClassifier	0.452830	0.363636	0.645
15	PassiveAggressiveClassifier	0.293103	0.515152	0.430
16	DNN	0.559322	0.500000	0.705

User Manual:

User Inputs through Google-form and Model Outputs through Email.

1st Method:

Manually entering all the 9 values from the lab report

Selecting Models

Entering Email-ID for report/models output

2nd Method: (Optional)

Clicking photo of the Lab report and uploading them

Selecting Models

Entering Email-ID for report/models output

Using OCR Technique, table/values are extracted from the image for model Input

Link to code:

https://colab.research.google.com/drive/1pg3axtCLtNTRYsh3lKoj-S_UTEJMCvUS?usp=sharing