

# **PREDICTING THE CONDITION OF TANZANIA WATER WELLS**

## **1. Business Understanding**

### **Problem Statement**

Lack of clean and potable water is a major issue in many communities across Tanzania. To address this issue, the Tanzanian Ministry of Water has installed several water wells across the country.

However, not all of these wells are functioning as intended, which results in a lack of access to clean water for communities.

The goal of this project is to build a predictive model that can accurately predict the condition of water wells in Tanzania based on data from Taarifa and the Tanzanian Ministry of Water to predict which pumps are functional, which need some repairs, and which don't work at all.

By doing so, the aim is to improve maintenance operations and ensure that clean and potable water is available to communities across Tanzania.

### **Research Question**

Which classifier model can accurately predict the condition of water wells in Tanzania?

### **Objectives**

#### **Main Objective**

To predict the condition of water wells in Tanzania to ensure that clean and potable water is available to communities across Tanzania.

#### **Specific Objectives**

- To understand the problem statement and the goal of the project
- To identify the variables that can impact the functionality of water wells
- To determine the target variable (functional, need repairs, or non-functional)

## **Metric of Success**

The model will be considered a success when both accuracy and f1 score are between 0.8 to 1.

## **2. Data Understanding**

### **Data Source**

Data is downloaded from "Pump it Up: Data Mining the Water Table" competition hosted by DrivenData

### **Data Description**

amount\_tsh - Total static head (amount water available to waterpoint)

date\_recorded - The date the row was entered

funder - Who funded the well

gps\_height - Altitude of the well

installer - Organization that installed the well

longitude - GPS coordinate

latitude - GPS coordinate

wpt\_name - Name of the waterpoint if there is one

num\_private -

basin - Geographic water basin

subvillage - Geographic location

region - Geographic location

region\_code - Geographic location (coded)

district\_code - Geographic location (coded)

lga - Geographic location

ward - Geographic location

population - Population around the well

public\_meeting - True/False

recorded\_by - Group entering this row of data

scheme\_management - Who operates the waterpoint

scheme\_name - Who operates the waterpoint

permit - If the waterpoint is permitted

construction\_year - Year the waterpoint was constructed

extraction\_type - The kind of extraction the waterpoint uses

extraction\_type\_group - The kind of extraction the waterpoint uses

extraction\_type\_class - The kind of extraction the waterpoint uses

management - How the waterpoint is managed

management\_group - How the waterpoint is managed

payment - What the water costs

payment\_type - What the water costs

water\_quality - The quality of the water

quality\_group - The quality of the water

quantity - The quantity of water

quantity\_group - The quantity of water

source - The source of the water

source\_type - The source of the water

source\_class - The source of the water

waterpoint\_type - The kind of waterpoint

waterpoint\_type\_group - The kind of waterpoint

### **3. Data Preparation**

- Uniformity - Handling Missing Values
- Consistency - Encoding Categorical Variables
- Completeness - Ensure the dataset has no missing values.
- Scaling
- Handling Class Imbalance
- Uniformity - Handling Missing Values

### **4. Modeling**

The Random Forest Classifier model has an accuracy of 0.8185 and an F1 score of 0.8110

In terms of accuracy, the random forest classifier has a higher accuracy score of 0.82 compared to the other 4 models, which had accuracy scores of 0.76 (gradient boosting), 0.76 (KNN), 0.75 (decision tree), and 0.66 (logistic regression).

In terms of the F1 score, the random forest classifier has a higher F1 score of 0.81 compared to the other 4 models, which had F1 scores of 0.74 (gradient boosting), 0.75 (KNN), 0.75 (decision tree), and 0.63 (logistic regression).

Based on these results, the random forest classifier appears to be performing the best among the 5 models.

Hyperparameter tuning and cross-validation is performed because they help to optimize the model performance and prevent overfitting or underfitting.

The randomized search cross-validation helps to perform an efficient search for the optimal hyperparameters and cross-validate the model, which leads to better model performance and robustness.

The accuracy of the model is 0.8221, which means that the model correctly predicted the class label of 82.2% of the test data instances.

The F1 score is 0.8135. A higher F1 score indicates a better performance of the model.

This tuned model has better performance metrics and hence a good model..

## **5. Challenging the Solution**

1. While the model has a high accuracy of 0.82, the precision and recall scores for the 'functional needs repair' class are lower compared to the other two classes. This could indicate that the model is not accurately predicting instances of this class and could potentially lead to misclassification of important information.
2. The imbalance in the number of instances for each class in the training data could lead to the model overfitting to the majority class and not accurately predicting instances of the minority class. To address these challenges, further refinement of the model, such as using methods to balance the class distribution in the training data, or using different algorithms, could be considered.
3. Random Search CV is computationally expensive and therefore not the most suitable model for large datasets.

## **6. Conclusion and Recommendations**

### **Conclusion**

- The best performing machine learning algorithm for this problem was the Random Forest Classifier with an accuracy of 82.21% and a weighted F1 score of 0.813
- The model could be further improved by incorporating more data especially for the functional needs repair class to handle imbalance for the classes

## **Recommendations**

- The Tanzania Ministry of Water should invest in better waterpoint types such communal standpipes and hand pumps
- The Tanzania Ministry of Water should ensure that the extraction type for the wells is mostly through gravity and handpump
- The Tanzania Ministry of Water should ensure that the gps height(altitude of the well) for most water points is high enough
- The Ministry of Water should also ensure that the people using the waterpoints pay either monthly, annually or per bucket to ensure that the wells are well maintained