## PREDICTING WATER WELL FUNCTIONALITY IN TANZANIA

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## PROJECT OVERVIEW

Why This Project Matters

Access to clean water is essential for life and development.

In Tanzania, many wells stop working due to poor maintenance or aging infrastructure.

This project uses data science to predict which water wells are **functional**, **non-functional**, or **need repair**, helping decision-makers prioritize resources and improve service delivery.

# BUSINESS UNDERSTANDING

The Business Challenge

Thousands of wells serve millions of people across Tanzania.

Many are nonfunctional, causing water shortages and repair delays.

Maintenance teams need a reliable way to identify high-risk wells early. Goal:
Provide a
predictive
model that helps
plan and prioritize
maintenance
efficiently.

# DATA UNDERSTANDING

About the Data

We used the **Tanzania Water Wells dataset** from the DrivenData "Pump It Up" competition.

It includes data on **59,000+ wells**, such as:

**Location:** region, district, coordinates

**Construction:** installer, year built, water quantity

Management: who runs the well, how users pay

**Target:** current condition — functional, needs repair, or non-functional

This data helps uncover what factors contribute most to well performance.

## MODELING APPROACH

How We Built the Model



We used **classification modeling**, which means the model predicts categories (e.g., "Functional" vs "Broken").



Two models were developed:



**Logistic Regression** – A simple model to set the baseline.



**Random Forest Classifier** – A more advanced model capturing complex relationships.



Both models were trained, tested, and compared to find the best performer.

## **RESULTS**

How Well Did the Models Perform?



**Random Forest** gave the highest accuracy and reliability.



It correctly predicted most wells' conditions and captured patterns missed by the simpler model.



#### Key takeaway:

The model can **accurately identify high-risk** wells, allowing proactive maintenance decisions.

#### **KEY INSIGHTS**

**What Factors Matter Most** 

The data revealed several important patterns:

Construction Year: newer wells tend to work better.

Water Quantity: low or dry wells often fail.

Management Type: wells managed by professional organizations are more reliable.

**Region:** some areas experience higher failure rates due to environmental conditions.

These insights can guide where to invest in repairs and new infrastructure.

## **RECOMMENDATIONS**

What We Recommend



PRIORITIZE
MAINTENANCE ON
OLDER WELLS AND THOSE
WITH LOW WATER
OUTPUT.



USE PREDICTIVE MODELING TO CREATE A NATIONAL MAINTENANCE PLAN.



**FOCUS FUNDING** ON HIGH-RISK REGIONS.



TRAIN LOCAL TEAMS
TO COLLECT BETTER
MAINTENANCE AND
MANAGEMENT DATA.

## **NEXT STEPS**

**What Comes Next** 



INTEGRATE THE MODEL INTO A **MOBILE OR WEB TOOL** FOR FIELD TEAMS.



COLLECT MORE **REAL- TIME WELL DATA** TO IMPROVE ACCURACY.



PARTNER WITH
GOVERNMENT AND NGOS
FOR NATIONWIDE
DEPLOYMENT.



CONTINUE REFINING THE MODEL FOR LONG-TERM SUSTAINABILITY.

# THANK YOU

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