Predictive Maintenance of Water Pumps in Tanzania

Enhancing Service Delivery Through Data Science, by Leonard Koyio

1. Introduction

Overview: This project focuses on using data science to enhance the maintenance of water pumps in Tanzania.

Importance: Functional water pumps are crucial for reliable water supply, especially in rural areas. **Goal**: Our aim is to predict the status of water pumps to ensure timely and efficient maintenance.

2. Business Understanding

Business Objective:

Assist the Tanzanian government in optimizing water pump maintenance by predicting pump functionality, ensuring consistent access to clean water for communities.

• Business Problem:

Inefficient allocation of maintenance resources due to a lack of predictive insights, leading to over-servicing functional pumps and under-servicing those needing repairs, increasing costs and water supply disruptions.

2. Business Understanding

Business Benefits

• Improved Resource Allocation:

Prioritize critical units, optimize resource use, and enhance service delivery by focusing efforts on pumps needing immediate attention.

• **Cost Savings**:Reduce repair costs and extend pump lifespan through preventative maintenance, leading to significant long-term savings.

Enhanced Water Access:

Ensure better access to clean water, vital for public health, by reducing non-functional pumps and improving water service reliability.

2. Business Understanding

Data-Driven Decision Making:

Foster a culture of data-driven strategies, improving efficiency in water pump maintenance and setting a precedent for other public services.

• Improved Service Delivery:

Targeted and efficient maintenance approach reduces downtime and ensures quicker repairs, enhancing public trust and satisfaction with government services.

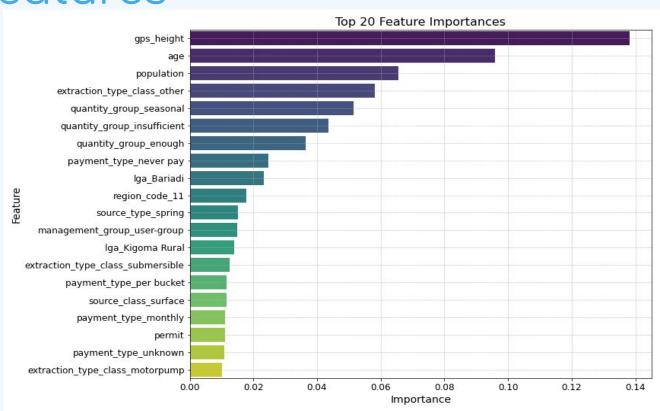
3. Data Understanding

Dataset Description: The dataset includes information on various features influencing the status of water pumps.

Key Features:

- **GPS Height**: Elevation of the pump location.
- **Age**: Age of the water pump.
- **Population**: Population served by the pump.
- **Extraction Type**: Type of extraction method used.
- Quantity Group: Water quantity availability (e.g., seasonal, insufficient, enough).

Visualization : top 20 model features



4. Exploratory Data Analysis (EDA)

• Data Cleaning:

- **Remove Redundant Columns**: Eliminated columns with no predictive value or columns with repeated data to streamline the dataset.
- **Handle Missing Values**: Filled zeros and NaNs with suitable data, such as median or mean values, to maintain data integrity.

Data Transformation:

- **Min-Max Scaling**: Applied Min-Max Scaling to normalize the range of independent variables, ensuring all features contribute equally to the model.
- **One-Hot Encoding**: Converted categorical variables into binary vectors to allow the model to interpret them effectively.

Feature Engineering:

 Age of the Pump: Calculated the age of each pump by subtracting the installation year from the current year, providing a crucial feature for predicting functionality.

5. Modeling

Baseline Models:

- Developed initial models using Logistic Regression and Decision Trees with default hyperparameters to establish a performance benchmark.
- Iteratively tuned models based on performance metrics.

Class Imbalance Handling:

- **Class Weights**: Adjusted weights to handle class imbalance within the model.
- **SMOTE**: Used Synthetic Minority Over-sampling Technique to balance class distribution by generating synthetic samples for the minority class.

5. Modeling

Model Regularization:

- **L1 Regularization**: Promotes sparsity by driving some coefficients to zero, improving model interpretability.
- **L2 Regularization**: Penalizes large coefficients to reduce overfitting, enhancing model generalizability.

Decision Tree Tuning:

- **Criteria**: Compared Gini and Entropy criteria to determine the best split quality measure.
- Max Depth Optimization: Tuned the maximum depth parameter to prevent overfitting and improve model performance.

6. Model Evaluation

Settled Model: Decision Tree trained on dataset without outliers.

- **Class Imbalance Handling**: SMOTE to balance class distribution.
- **Criterion**: Gini impurity criterion.
- Max Depth: None specified.

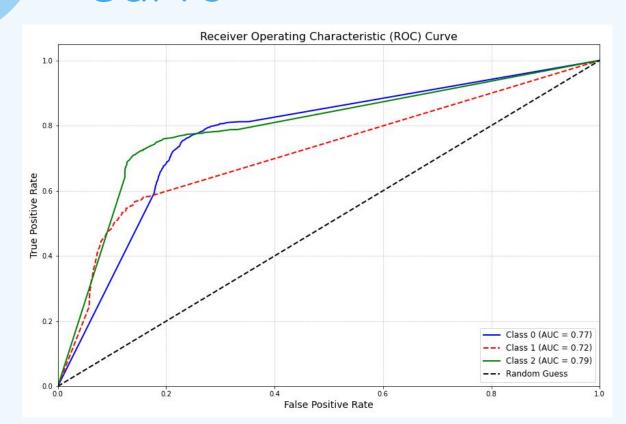
Metrics:

- Weighted Recall: 0.7128
 - Indicates that the model correctly identifies 71.28% of instances, considering class distribution, which is critical for ensuring non-functional and repair-needed pumps are detected.
- **Weighted Precision**: 0.7443
 - Shows that 74.43% of the positive predictions are correct, reflecting high accuracy in identifying pumps needing attention.

6. Model Evaluation

- Weighted F1 Score: 0.7253
 - Balances precision and recall, ensuring the model is effective in both identifying and accurately predicting pump statuses.
- **ROC AUC Score**: 0.7626
 - Measures overall performance; a score of 0.7626 indicates a good balance between sensitivity and specificity, which is suitable for our objective of predictive maintenance.

7. Model Visualizations : a. ROC Curve

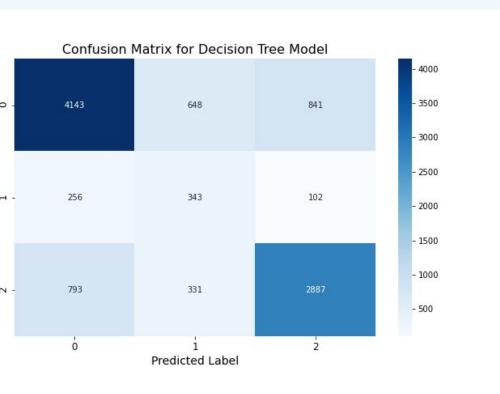


Class 2 (Non-Functional): High AUC of 0.79 ensures accurate identification of out-of-service pumps, crucial for timely repairs.

Class 1 (Needs Repair): AUC of 0.72 indicates good detection of pumps needing maintenance, preventing further issues.

Class 0 (Functional): AUC of 0.77 helps confirm operational pumps, reducing unnecessary checks and optimizing resource use.

7. Model Visualizations : b. Confusion Matrix



Key

0 - Functional

1- Functional Needs Repair

2- Non- Functional The model is well-balanced, with no strong bias towards any class, ensuring fair consideration of all pump statuses.

8. Predictive Recommendations

1. Investigate Zanzibar South (Region Code 11):

Address regional-specific challenges affecting pump performance to enhance maintenance efforts.

2. Strengthen Bariadi's Local Government Authority:

Support Bariadi's LGA with training, resources, and funding to better manage and maintain pumps.

3. Tailor Maintenance by GPS Height:

Customize maintenance schedules based on pump elevation to address altitude-related issues.

4. Focus on Older Pumps:

Implement proactive maintenance strategies for older pumps to extend their lifespan and prevent failures

8. Predictive Recommendations

5. Improve Seasonal and Insufficient Water Supply:

Enhance maintenance practices for pumps with seasonal or insufficient water supply to ensure consistent water availability.

6. Train Communities for Handpump Maintenance:

Provide training for communities to effectively maintain handpumps, reducing failure rates.

7. Promote Regular Payment for Water Services:

Encourage consistent payment for water services to support sustainable maintenance funding.

8. Allocate Resources Based on Population:

Prioritize maintenance resources for areas with higher populations to manage increased wear and tear.

8. Predictive Recommendations

Model Utility:

- **Most Useful**: In rural or remote areas where quick maintenance decisions are crucial. Helps prioritize repairs and maintenance.
- Less Effective: In rapidly changing conditions or where historical data is outdated. Suggestions for Improvement:
 - Data Accuracy: Ensure accurate and up-to-date data on pump conditions and maintenance.
 - **Feature Engineering**: Add features related to usage patterns, maintenance history, and environmental conditions.

9. Project Impact

Improved Maintenance Efficiency: Enables timely repairs and consistent water supply.

Cost-Effective Resource Allocation: Optimizes budget use and reduces unnecessary expenses.

Scalability and Adaptability: Easily adapts to include more features or updated data.

10. Conclusion

Project Success:

- Developed a robust predictive model that categorizes water pumps as functional, needing repair, or non-functional using data from Taarifa and the Tanzanian Ministry of Water.
- Provided predictive recommendations to improve water pump maintenance based on the model
- Directly addresses the need for efficient water pump maintenance in Tanzania.

Model Advantages:

- Accuracy: The Decision Tree Classifier delivers reliable predictions, essential for timely maintenance decisions.
- Efficiency: Enhances maintenance scheduling and reduces downtime, ensuring consistent water supply.

10. Conclusion

- **Cost-Effectiveness**: Optimizes resource allocation and minimizes expenses by identifying priority pumps for repair.
- **Scalability**: Easily adaptable to incorporate new data or features, supporting ongoing infrastructure management.

Stakeholder Value:

- **For the Government**: A critical tool for optimizing water infrastructure management and aligning with national goals for improved water access.
- **For Citizens**: Ensures reliable access to clean water, contributing to enhanced community well-being.