TANZANIAN WATER WELLS

WELLS

OVERVIEW

Driven Data competition > Pump it up: Data Mining the Water Table -2015

Reopened and ongoing >6 more months

Taarifa: Winner of the London Water Hackathon -2011

Deployed in Tanzania -2012

Prediction of functionality -for better use of resources

DATA

59400 Wells

39 Features:

- Location
- Logistics
- Pump
- Water

Many similar hierarchical features, e.g.: extraction_type_class < extraction_type_group < extraction_type

Class Distribution

Functional: 54%

Need's repair: 7%

Non-Functional: 38%

EXPLORATORY DATA ANALYSIS

Categorical Data:

30 out of 39 features

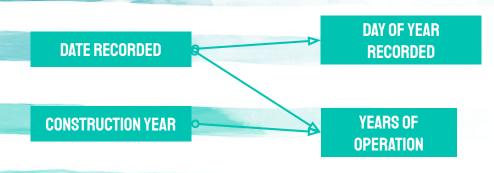
Some with thousands of categories

Dropped Features
Trimmed Categories
One Hot Encoded

Dropped Features
Target encoding

Catboost

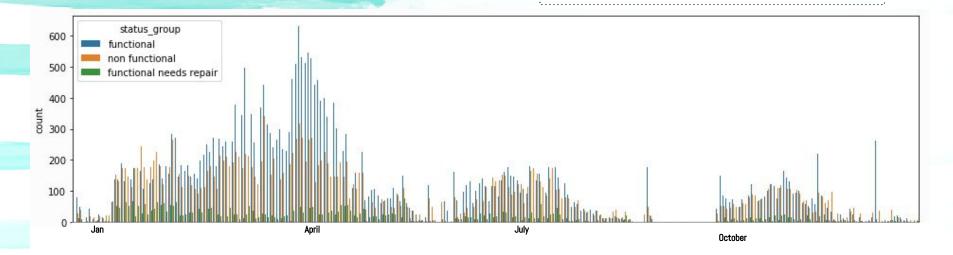
EXPLORATORY DATA ANALYSIS



Climate:

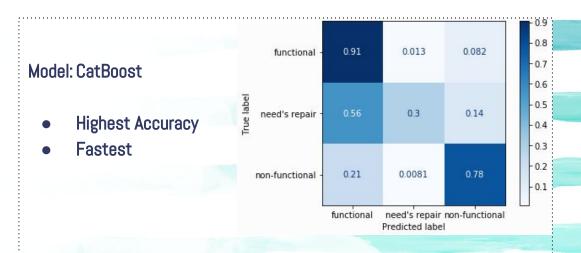
'Long Rains': March, April, May

Higher ratio of well pumps found functional during the rainy season.



MODELS

| Model | Accuracy |
|---------------------------|----------|
| Logistic Regression | 0.71 |
| K Nearest Neighbors | 0.78 |
| Naive Bayes | 0.69 |
| Decision Tree | 0.78 |
| Random Forest | 0.73 |
| Adaboost | 0.74 |
| Gradient Boosting | 0.80 |
| Support Vector Classifier | 0.80 |
| XGBoost | 0.74 |
| CatBoost | 0.81 |



Poor predictor of 'need's repair' -minority class(7%)

mostly predicted as 'functional'

CONCLUSION

CatBoost model -81% accurate

Important features: mostly location based

- Same source?
- Recorded at the same time?

More data:

- Seasonally functional -as a class
- When did a well pump become non-functional

FUTURE WORK

- SMOTE on the smallest class
- More hyperparameter tuning in XGBoost and SVC
- Model bagging

- Use CatBoost for imputing
- Try H20 algorithm

THANK YOU LHANK AON

Thanks also to:

- Taarifa
- Tanzanian Ministry of Water
- DrivenData

Matthew Andrews

GitHub Repo:

https://github.com/Maltanno/Phase3_Project