# ML COURSE PROJECT

Google Analytics Customer Revenue Prediction

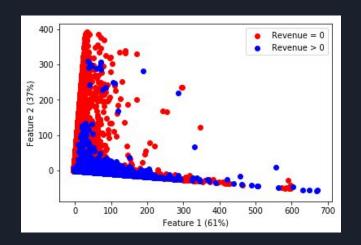
Nihesh Anderson Shravika Mittal Pragya Prakash

### Problem Statement

- Kaggle challenge <u>Google Analytics Customer Revenue Prediction</u>
- Given the details of customers hitting GStore website and their transaction details, the objective is to predict the log revenue per user.
- Evaluation Metric: RMSE Error of all users, grouped by unique FullVisitorIDs

### Dataset

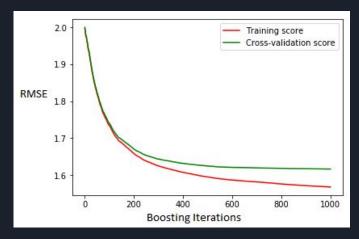
- 1. Almost 99% of the transactions do not generate revenue
  - a. Transactions generating revenue 11515
  - b. Transactions not generating revenue 892138
- 2. One user can have multiple transaction entries
  - a. Unique users 714167 (9996 users generate revenue)
- 3. Data visualization using PCA
  - a. Number of features reduced to 2 using PCA
  - b. Some overlap can be seen between the data points generating revenue and those which do not generate revenue



<sup>\*\*</sup> More insights in backup slides

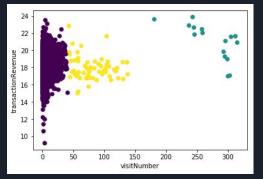
### Approaches Used

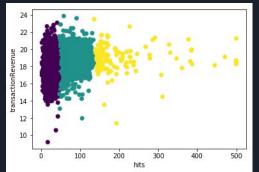
- Baseline Model LASSO
- Robust SVM
- Dense Neural Network
- Random Forest Regressor
- Light Gradient Boosting Machines (LGBM)

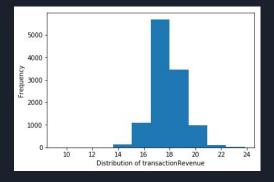


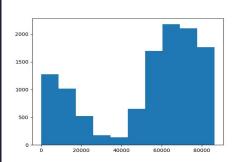
Plot of training and validation rmse scores at intervals of 50 boosting rounds of our best model: Light GBM

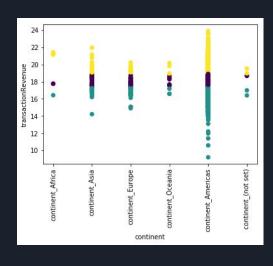
## Exploratory Data Analysis









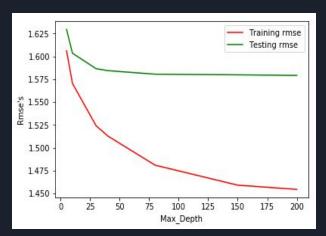


### Results

- Interim evaluation result 1.7000 (RMSE on test data)
- Kaggle benchmark 1.4500

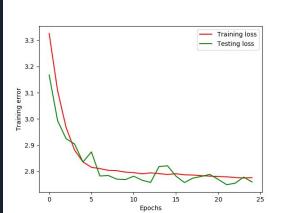
| Model Used              | RMSE on Train data | RMSE on Test data |  |
|-------------------------|--------------------|-------------------|--|
| Baseline - LASSO        | 1.8159             | 1.8273            |  |
| SVM                     | 1.6912             | 1.7400            |  |
| Random Forest Regressor | 1.6114             | 1.6390            |  |
| Neural Network (MLP)    | 1.6029             | 1.6282            |  |
| Light GBM               | 1.5265             | 1.5812            |  |

## Model Analysis and Parameter Tuning



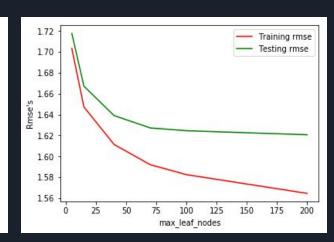


- Max\_leaf\_nodes = 40, max\_depth = 15
- More depth leads to more accuracy on train set, but poor generalisation, i.e., overfitting
- Max-Depth ensures good fit.



#### **Neural Network Training**

- Trained with high batch size
- Training was stopped at 12 epochs
- Early stopping ensures that the model doesn't overfit.



#### Random Forest Training

- Max\_leaf\_nodes = 40 (GridsearchCV)
- More leaf nodes leads to more complexity causing overfitting.
- Above trend can be seen in the graph.

## Ablative Analysis - Test Error

| Component      | LightGBM | Neural Network | Random Forest |
|----------------|----------|----------------|---------------|
| Overall System | 1.5812   | 1.6283         | 1.6390        |
| isMobile       | 1.5855   | 1.6545         | 1.6381        |
| hits           | 1.6133   | 1.6602         | 1.6420        |
| continent      | 1.6142   | 1.7000         | 1.6425        |
| pageviews      | 1.9097   | 1.9476         | 1.9181        |

## Learning Curves





Light GBM learning curve

Neural Network learning curve

Random Forest learning curve

### Individual Contribution

#### Nihesh Anderson

- Data processing (Parsing unstructured data, encoding categorical data)
- Exploratory Data Analysis
- Trained Lasso Regression Baseline
- Tuning Neural Network Architecture using Keras

#### Shravika Mittal

- Exploratory Data Analysis
- Trained Decision Trees and Random Forests using scikit learn
- Feature importance from Random Forest used for feature engineering

#### Pragya Prakash

- Exploratory Data Analysis
- Trained SVM using scikit learn
- Trained LightGBM using lightgbm library