### Digital Green Crop Yield Estimate Project

**DTSA 5509 November 10, 2024** 

#### **Data Challenge Introduction**

The contest is sponsored by Digital Green (https://digitalgreen.org/).

The data challenge is to create a machine learning solution to predict the crop yield per acre of rice or wheat crops in India.

The data was collected through a survey conducted across multiple regions in India, which consists a variety of features, e.g., the type and amount of fertilizers used, the quantity of seedlings planted, irrigation methods, and etc.

- 5000 data records
- 43 features
- training/test data ratio: 75:25

The data challenge is available: https://zindi.africa/competitions/digital-green-crop-yield-estimate-challenge

#### **Analysis Proposal**

- 1. Predict the crop yields per acre using different models
  - Linear Regression
  - Tree-based Regression:

Extra Trees Algorithm
Lightweight Gradient Boosting Machine (LightGBM)
(and Ensemble model: the mixture of Extra Tree and LightGBM)

2. Evaluate the root mean square error (RMSE) across different models

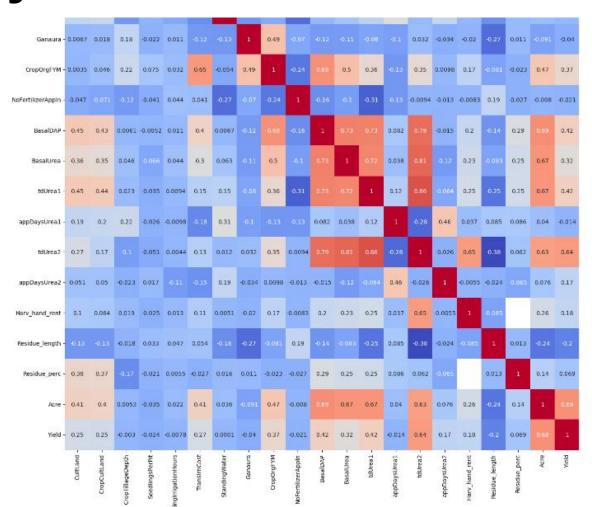
#### **Exploartory Data Analysis**

**Response: Yield** 

Factors: all other variables collected

**Data Cleaning -** check collinearity

- Dropped the categorical variables, focus on the effects from numerical variables.
- Removed the highly correlated numerical variables
- Deleted the records with missing Yield value

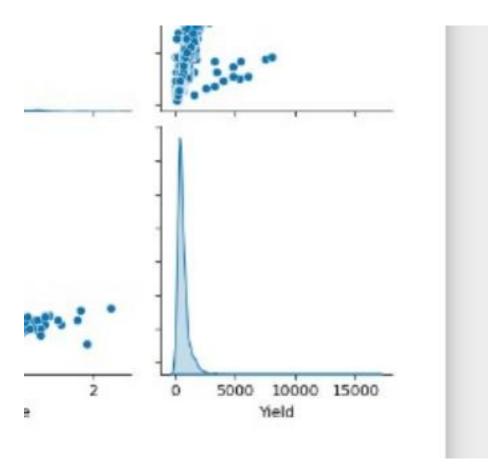


### **Exploartory Data Analysis**

#### **Response Variable Derivation**

- Log transformed Yield as it is left-skewed distributed
- Derived new Yield variable :

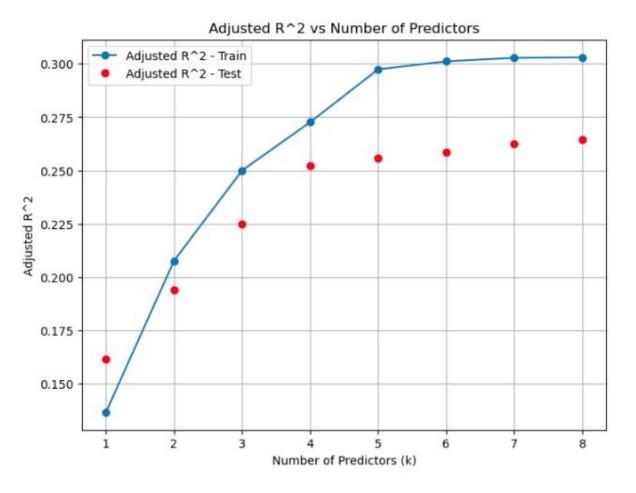
Yield per acre = Yield / Acre



## **Outputs - Multiple Linear Regression**

- Forward feature selection
- Good prediction with 5 or 6 features.
- Final Model:

| Variable Name    | Variable Description                                                            |
|------------------|---------------------------------------------------------------------------------|
| BasalDAP         | Amount of DAP(in kgs)applied during land preparation                            |
| CropCultLand     | Area of land under cultivation                                                  |
| TransIrriCost    | Cost of irrigation during transplantation                                       |
| Residue_length   | Length of the residue left after harvesting                                     |
| Harv_hand_rent   | If labours were used or harvesting machine hired, what was the rent (in rupees) |
| TransIrriCost    | Cost of irrigation during transplantation                                       |
| CropOrgFYM       | Amount of FYM (Farm yard manure) organic fertilizer used (in Quintals)          |
| CropTillageDepth | Depth of the tillage                                                            |

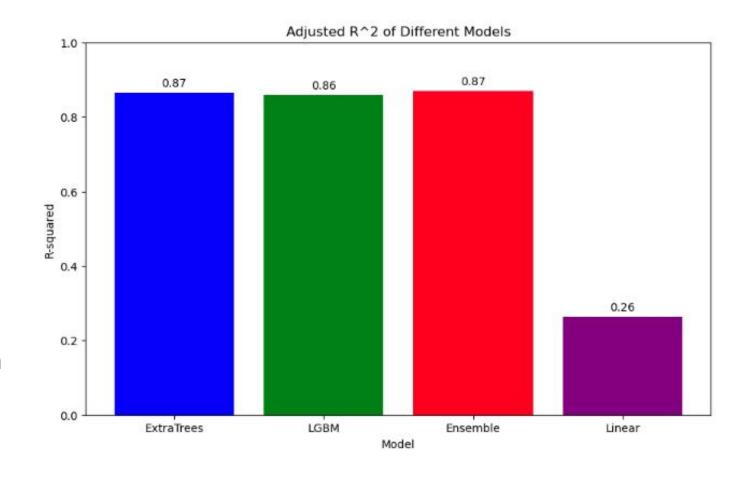


#### **Outputs - Tree-based Models**

Compare the model performance using adjusted R<sup>2</sup> using testing datat

- Extra Tree
- LightGBM
- Ensemble model is the average of Extra Tree and LightGBM modele

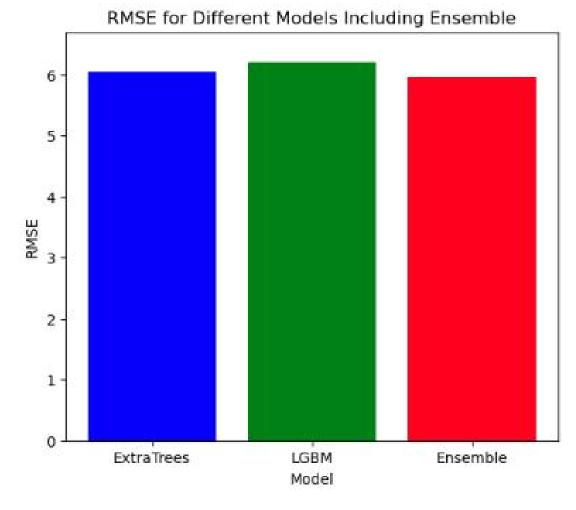
All the three tree-based models have much higher adjusted R<sup>2</sup> values.



#### **Outputs - Tree-based Models**

Comparison of RMSE

The three tree-based models have similar RMSE.



#### **Conclusion and Discussion**

- In the final prediction model, I select the ensemble model, which is a mixture of Extra Tree and LightGBM
- The multiple linear regression is not recommended for real-world data, but could be a good practice for exploratory analysis.
- Limitations:
  - Categorical features are not taken into consideration.
  - Overfitting vs. Underfitting

#### Resources

1. Extra Tree and LightGBM Coding Resources

https://scikit-learn.org/dev/modules/generated/sklearn.ensemble.ExtraTreesRegressor.html https://lightgbm.readthedocs.io/en/stable/Python-Intro.html

2. Data Source

https://zindi.africa/competitions/digital-green-crop-yield-estimate-challenge

3. My GitHub

https://github.com/lzheng01/Digital-Green-Crop-Yield-Estimate-Project

# Thank You!