

Machine Learning • Data Analysis • Feature Engineering

By Asheesh Kumar



# Objective

• To classify Instagram accounts as fake, spammer, or genuine using machine learning models, exploratory data analysis (EDA), and feature engineering.

### Dataset Description

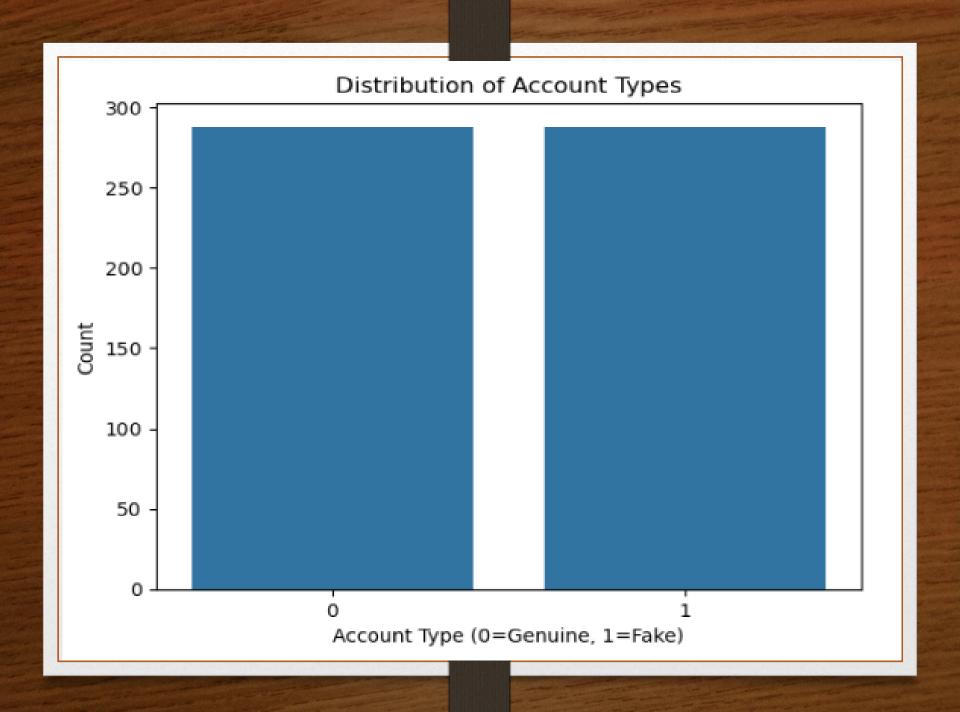
- train.csv: Includes account features and labels (fake = 1, genuine = 0)
- test.csv: Unlabeled Instagram accounts to classify.
- Features include: description\_length, #followers, #follows, #posts.

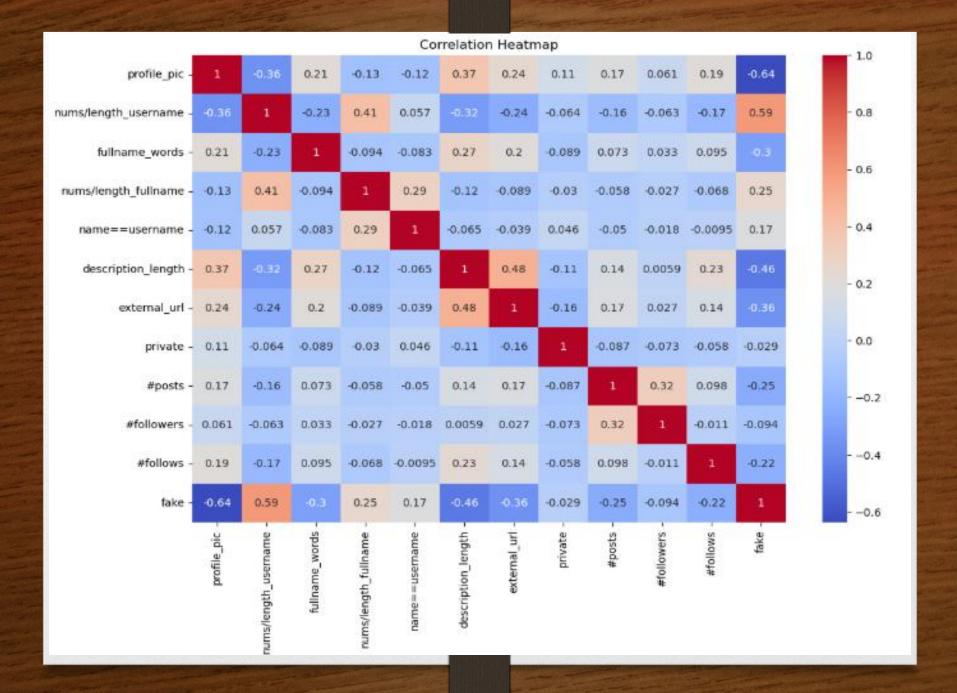
### Libraries & Tools

- Python (Pandas, NumPy, Scikit-learn)
- XGBoost, Matplotlib, Seaborn
- SHAP (Explainability)
- Jupyter Notebook, VS Code

### EDA Highlights

- Data imbalance: more genuine accounts than fake.
- Visualizations: count plots, heatmaps, boxplots.
- Highly correlated features: #followers, #follows, #posts.



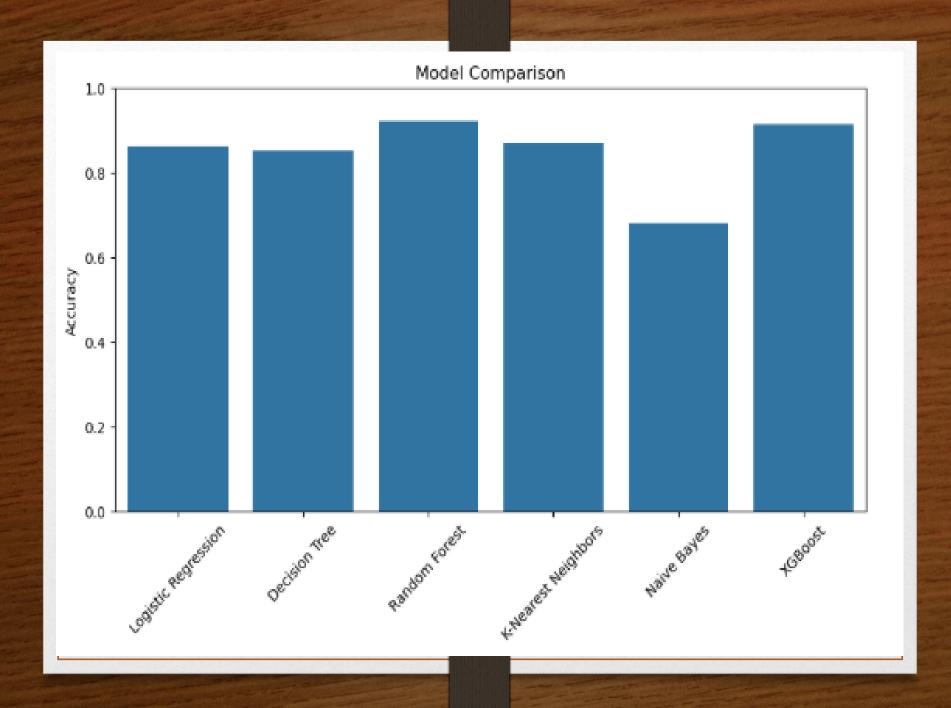


### Preprocessing Steps

- Cleaned and renamed columns.
- Label encoding of categorical features.
- Standardized numerical features.
- Split dataset: 80% train, 20% validation.

#### Model Evaluation

- Models Trained:
- Logistic Regression
- Decision Tree
- Random Forest
- KNN
- Naive Bayes
- XGBoost
- Best Accuracy: XGBoost



### Feature Importance

- Top Features from SHAP:
- #followers
- #follows
- description\_length
- #posts

## Predictions & Output

- Test data classified using best model.
- Predictions saved to final\_predictions.csv

Model: Logistic Regression Accuracy: 0.8620689655172413 Confusion Matrix: [[60 3] [13 40]] Classification Report: precision recall f1-score support 0.82 0.95 0.88 63 0.930.75 0.83 53 0.86 116 accuracy 0.86 0.88 0.85 116 macro avg 0.86 0.86 weighted avg 0.87116

### Conclusion & Future Scope

- XGBoost performed best on validation data.
- Future Improvements:
- SMOTE for imbalance
- Hyperparameter tuning
- Stacking/Ensembling

## Project By

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