



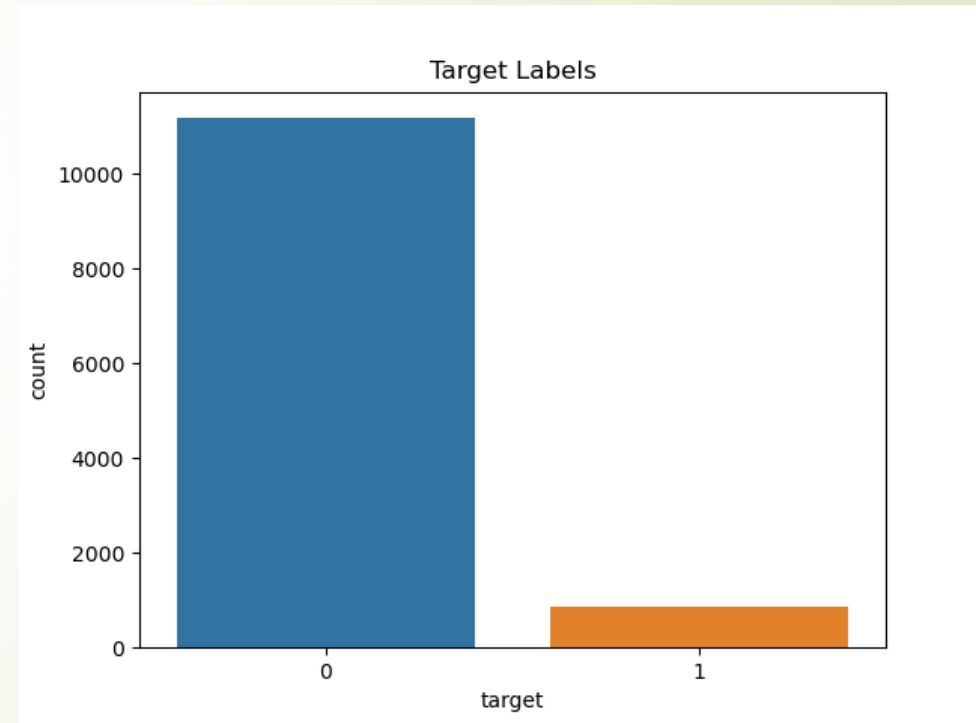
# Receipt Matching Data Science Challenge

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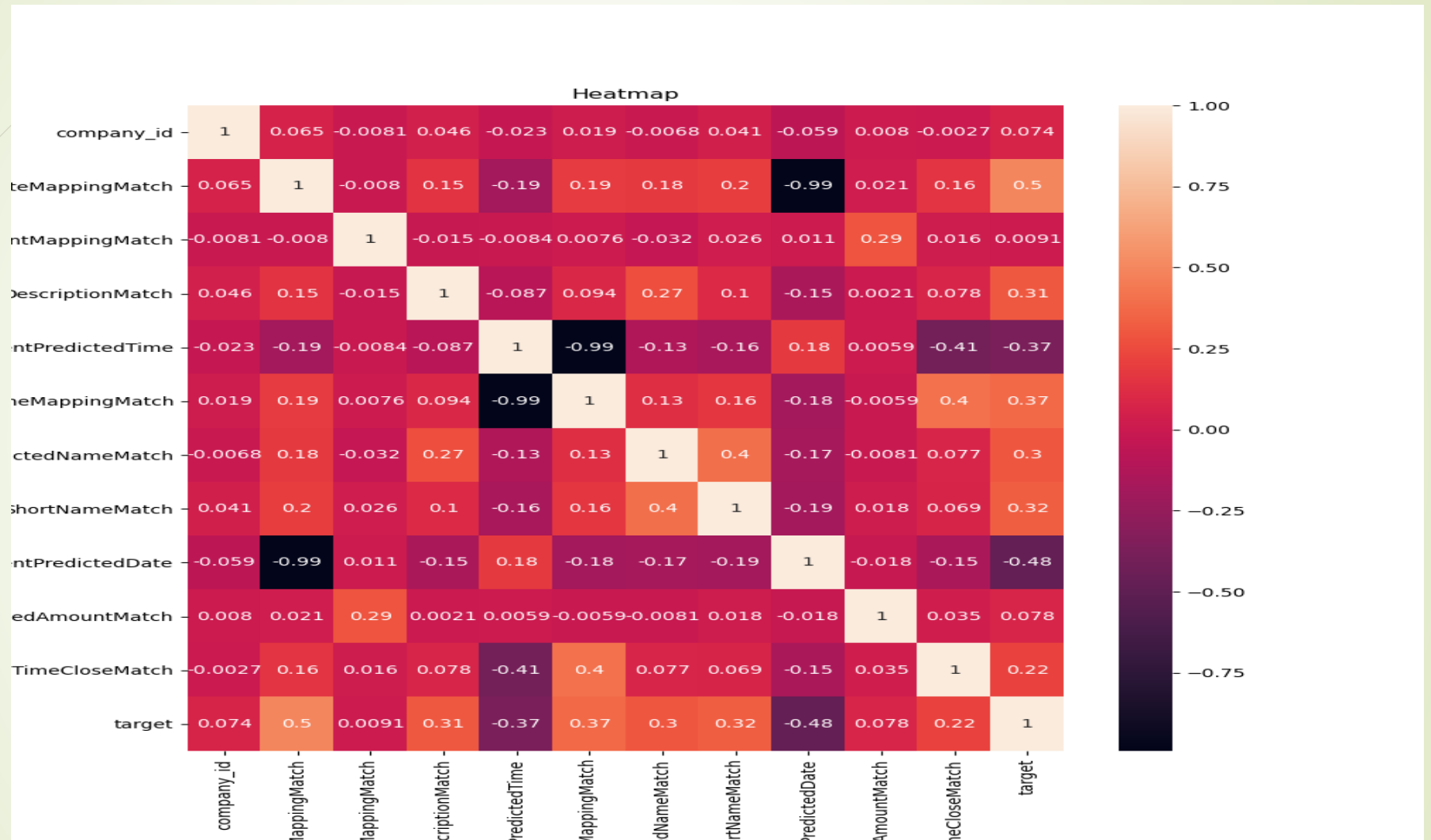


# Goals & Dataset Description

- Automatically match the receipt images with the transactions associated.
- In the app when the customer takes a picture of a receipt, the app provides a list of transactions likely to match the receipt, goal is map it to the correct transactions from the list.
- Highly imbalanced dataset, with only 857 correct matches and 11177 incorrect matches.
- Correct matches are defined if `matched_transaction_id` equals `feature_transaction_id`.

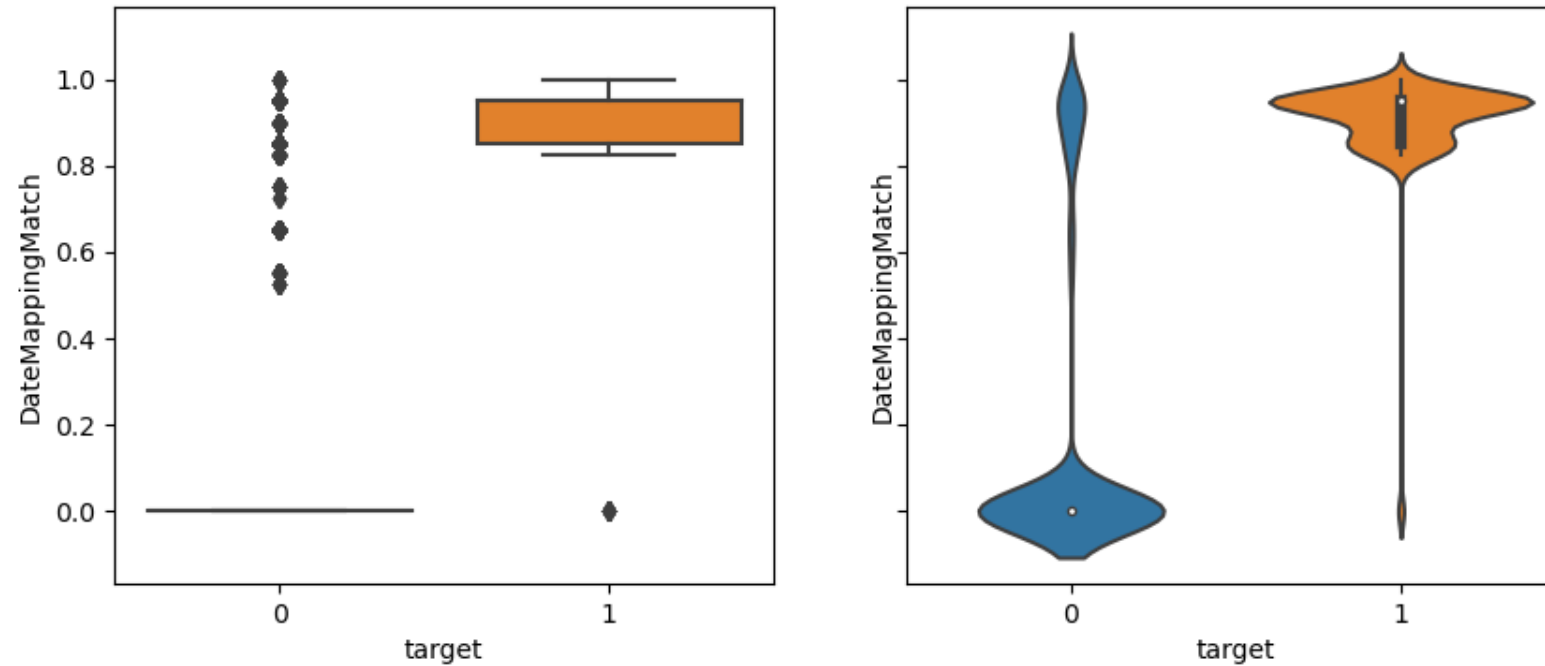


# EDA



**DifferentPredictedTime, DifferentPredictedDate can be dropped as they are negatively correlated with respect to target.**

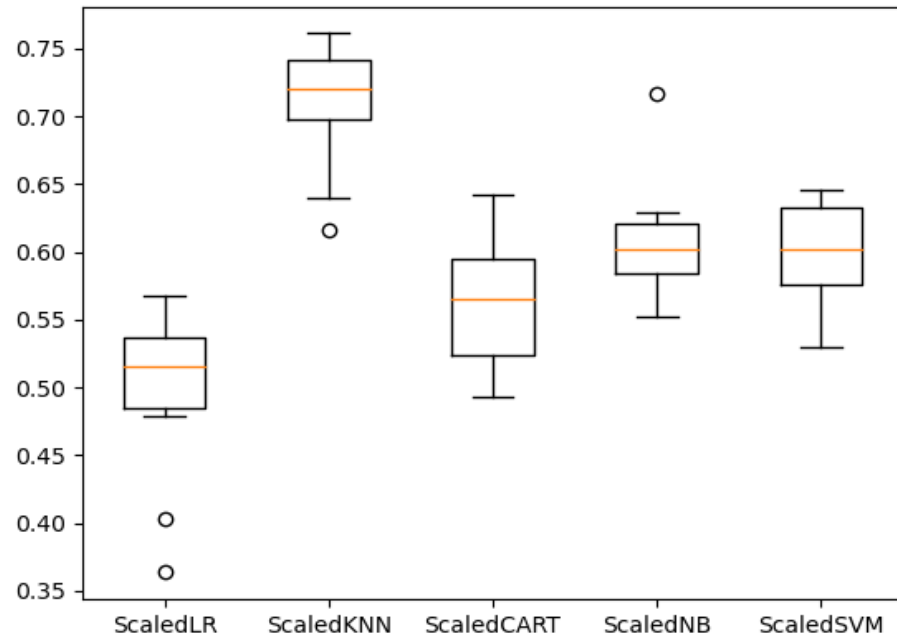
## EDA – Contd..



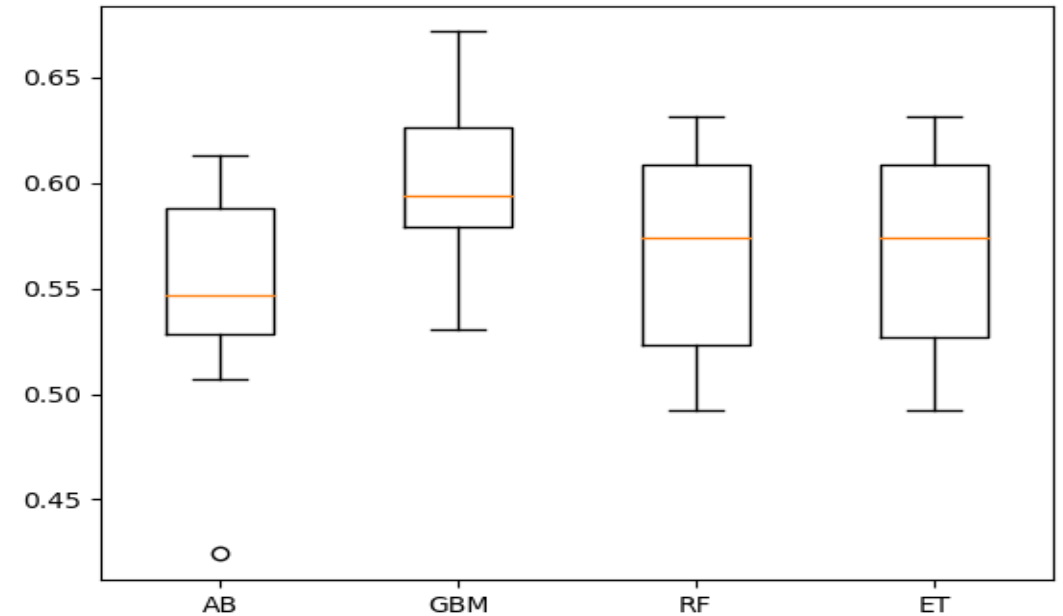
**Boxplot & violinplot for DateMappingMatch with respect to target.**

# Comparison Of Different Algorithms

Scaled Algorithm Comparison

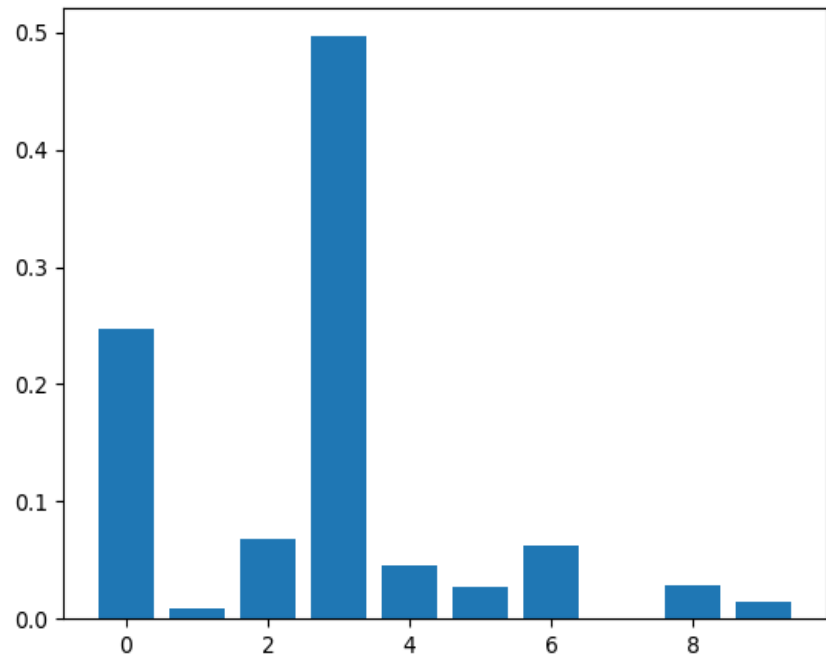


Ensemble Algorithm Comparison

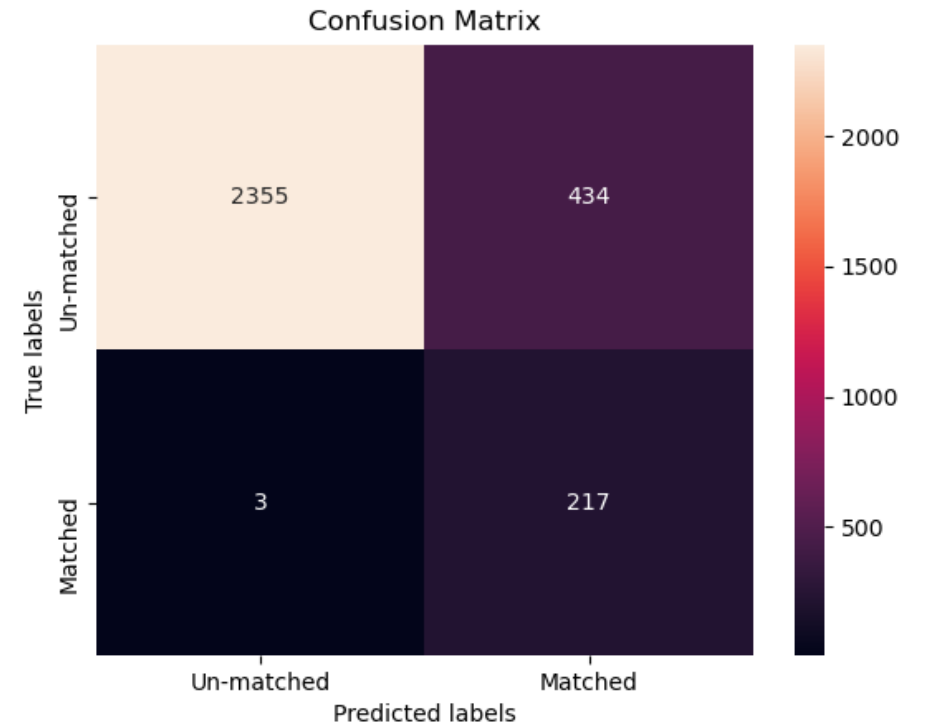


KNN performs better than the others...

# Use of XGBOOST



Feature importance using Xboost Classifier.



Using XBOOST gives 0.99 recall.



### ► **Future steps:**

- Hyperparameter optimization of XGBOOST, to improve the metrics further.
- Use of Oversampling/ undersampling techniques to balance the dataset.
- Use of deep learning based approaches for larger dataset.

### ► **Conclusion:**

- Recall is chosen as the main evaluation metric to consider.
- Hyper-Parameter optimization for Xgboost and others might not significantly improve the results as the dataset is imbalanced.
- Dropping negative correlated features/ outliers didnt produce better results.
- Data transformation might be needed after understanding of each of the column features.
- Dimensionality reduction / DL based approaches can be explored.



*Thank You!*