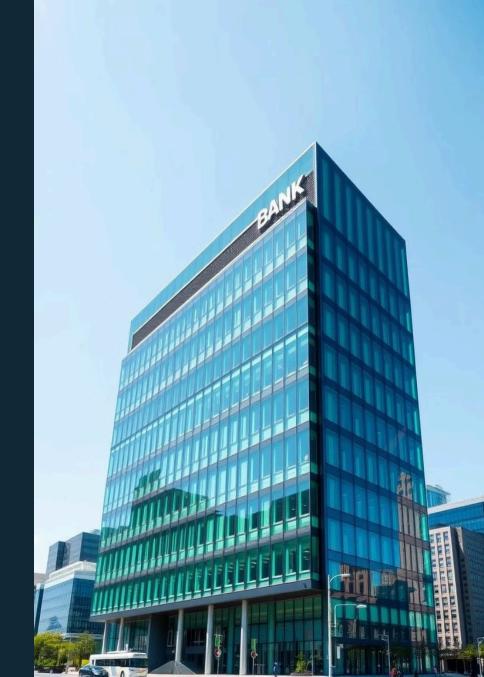
Bank Customer Complaint Classification

A Natural Language Processing (NLP) Project

November 2024

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Outline

3 **Project Overview** Business Data Understanding & Understanding Preparation 6 Conclusion & Modeling & Evaluation Deployment & Application Recommendations

Project Overview

Goal

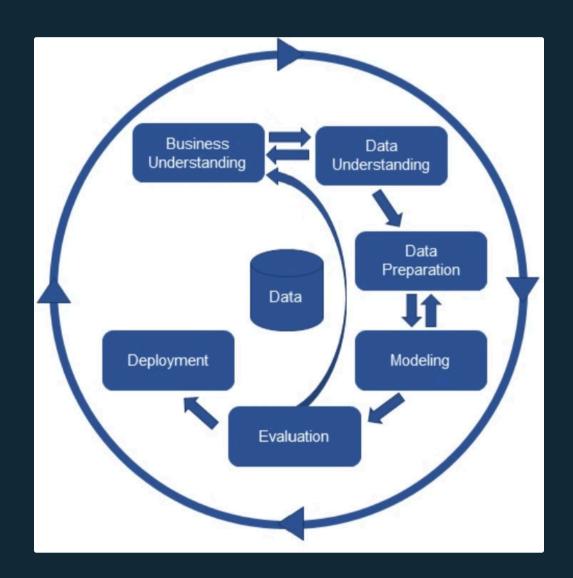
Simplify Customer complaint submission & handling.

Objectives

- Train an NLP model for accurate classification.
- 2. Reduce complaint submission time.
- 3. Enhance bank responsiveness.

Methodology

Cross-Industry Standard Process for Data Mining (CRISP-DM).



Business Understanding

Problem Statement

Customers face frustration due to complex complaint submission processes.

Root Causes

Complex navigation, inefficient chatbots, lack of personalization and limited data use.

Key Stakeholders

- 1. Customers and,
- 2. Customer support teams.



Data Understanding & Preparation

Dataset Source

US Consumer Complaints

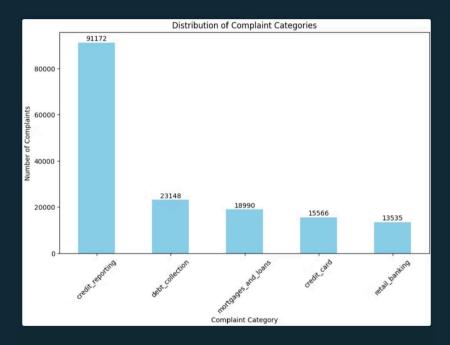
Dataset from <u>Kaggle</u>.

Complaint Categories

Five (5): Credit reporting; debt collection; mortgages, credit cards; and retail banking.



162,400 records, imbalanced data, and varying narrative lengths.



The distribution of Complaint Categories shows a class imbalance, with the credit_reporting category significantly more represented than others.

Data Understanding & Preparation

Exploration

Handled missing values, retained duplicates, addressed class imbalance.

Pre-processing

Standardized text, tokenization, stop word removal, and lemmatization.

Data Transformation

Applied TF-IDF and MinMax scaling. SMOTE also applied later for model improvement.

```
chargeonedispute company see uniformed someone way website credit limit requested informed statement merchant name offer Capital Called Credit reports and account claim as keep called credit reports well advised paid account closed balance paid account closed balance paid account closed balance advised put Customer went Service paid account closed balance account c
```

Word cloud highlighting specific terms associated with common issues for each product. Shows that customer complaints are centered around distinct themes based on the product type.

Data Understanding & Preparation

SMOTE Class Balancing

Synthetic Minority Oversampling Technique (SMOTE)

- Perfectly balanced class distribution after applying SMOTE.
- Each product category having an equal count of samples (approximately 73,000).
- This balanced distribution helps the model learn patterns more evenly across classes.





Models

Baseline: MNB, SVM,

Logistic Regression,

Random Forest.

Improved: Tuned

Random Forest,

ExtraTrees, BERT.



Evaluation Metrics

Classification report,

Confusion Matix, AUC-

ROC curve with focus

on Macro F1 Score

supported by

Weighted F1-score.



Model Selection

BERT seleted for its balanced performance and suitability for NLP tasks.



BERT Model Performance

Metrics Report:

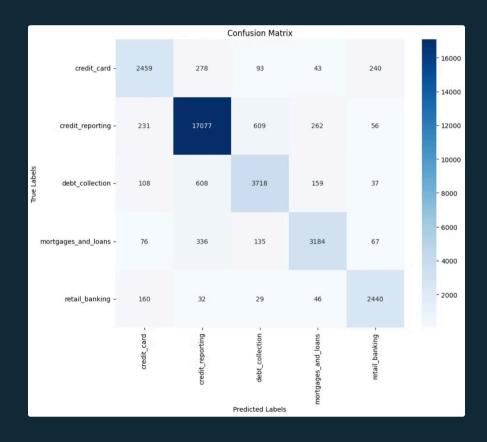
- Solid performance.
- The Macro F1 of 0.85 shows consistent results across classes.
- Weighted F1 0.89 and Accuracy of 89%.

Classification Report	:			
	precision	recall	f1-score	support
credit_card	0.81	0.79	0.80	3113
credit_reporting	0.93	0.94	0.93	18235
debt_collection	0.81	0.80	0.81	4630
mortgages_and_loans	0.86	0.84	0.85	3798
retail_banking	0.86	0.90	0.88	2707
42/2004/700/80070				
accuracy			0.89	32483
macro avg	0.85	0.85	0.85	32483
weighted avg	0.89	0.89	0.89	32483

BERT Model Performance

Confusion Matrix:

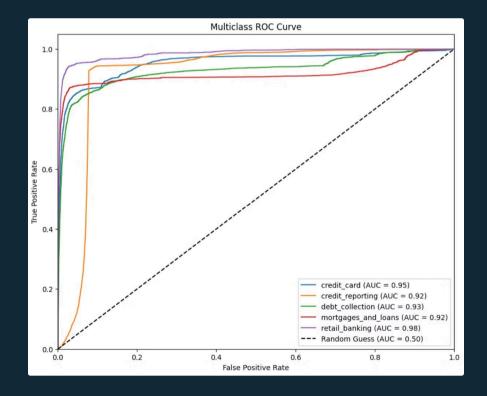
- Indicates strong classification performance.
- Some misclassification occurred between similar financial categories.
- Shows good overall differentiation.



BERT Model Performance

AUC-ROC Curve:

- Demonstrates high discriminative power across all complaint categories.
- AUC values ranging from **0.92 to 0.98**.
- Overall reflects a strong ability of the model to distinguish between classes.



Deployment & Application

User-Friendly Interface

Built with Streamlit & deployed on <u>huggingface</u>.

Model Integration

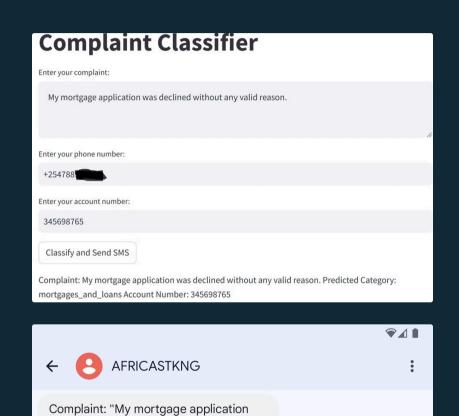
BERT classifies complaints in real-time.

Automated Notifications

Integrated with Africastalking's SMS API, sending classified notification.

Challenges

SMS limited to Airtel and Telkom; Email notification slow.



Snapshot of the system simlified user interface and SMS notification received

was declined without any valid reason."

Predicted Category:

mortgages_and_loans

Conclusion & Recommendations

- Insight & Impact
 - BERT model with robust classification accuracy trained.
 - Simple complaint submission system deployed.
 - 3. Prompt notification for enhanced responsiveness.

- Future Improvements
 - Feedback loops.
 - Category optimization.
 - Expanded notification channels.





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Thank you.