

SAFARICOM TWEETS CLASSIFICATION



TEAM ALPHA PRESENTATION



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TABLE OF CONTENTS





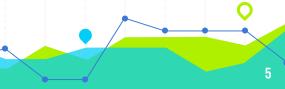
OVERVIEW

The advancement of technology has enhanced communication abilities with the use of social media in business and other sectors. Most companies have therefore resolved to creating accounts that specifically deal with customer care queries. Attending to these queries promptly is therefore at the core of most successful businesses as developing good customer interaction tools is key to increasing customer satisfaction.



PROBLEM STATEMENT

In recent years, a number of companies have migrated to using social media for customer service and customer support, with companies having specific accounts dedicated to customer care queries. This has led to the problem of classifying said complaints or services according to the department that needs to work on them and generate tickets for the services.



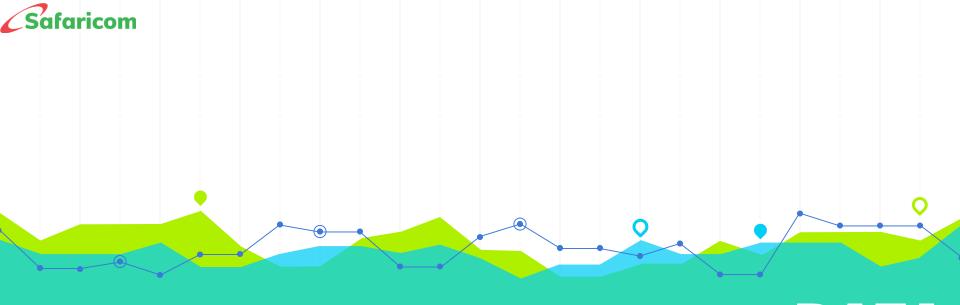


SPECIFIC OBJECTIVES

- To identify the departments in which the complaints fall in
- To improve customer service efficiency
- To improve branding and customer relations
- To accelerate customer target marketing



For the project success criteria we will be aiming for an F1 score of 70% to determine the efficiency and usability of our model.



DATA UNDERSTANDING

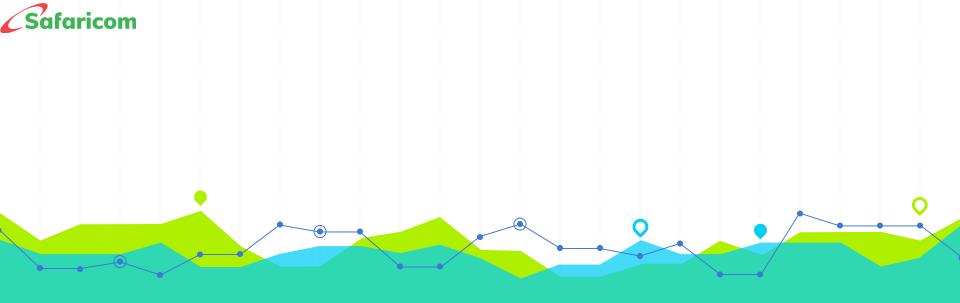


Tweets!



DATA DESCRIPTION

Column	Description				
Datetime	This refers to the date of the tweet				
Tweet Id	This refers to the unique id of the tweet				
Text	This refers to the actual text in the tweet				
Username	This refers to the name of the user who commented				
Like Count	This refers to the number of likes a tweet receives				



DATA PREPARATION



DATA CLEANING

@SAFARICOM_CARE

Removing all the tweets that are from Safaricom_Care.



REMOVING NOISE

Removing: links, usernames, hashtags, punctuations, emojis, stopwords and numbers

Time of day the tweet was sent, either morning, afternoon or evening.

TIME_OF_DAY



Creating labels for our data using key words manually

LABELS

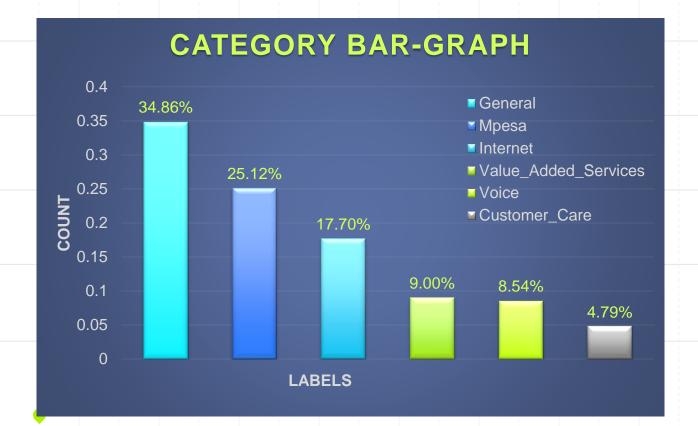


EXPLORATORY DATA ANALYSIS

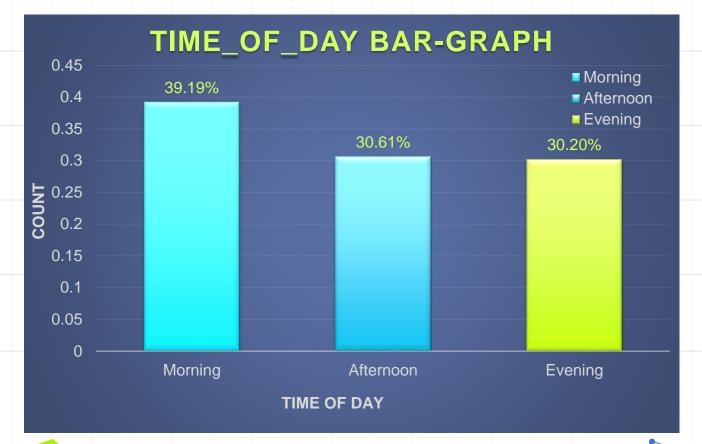
Univariate Analysis

Bivariate Analysis



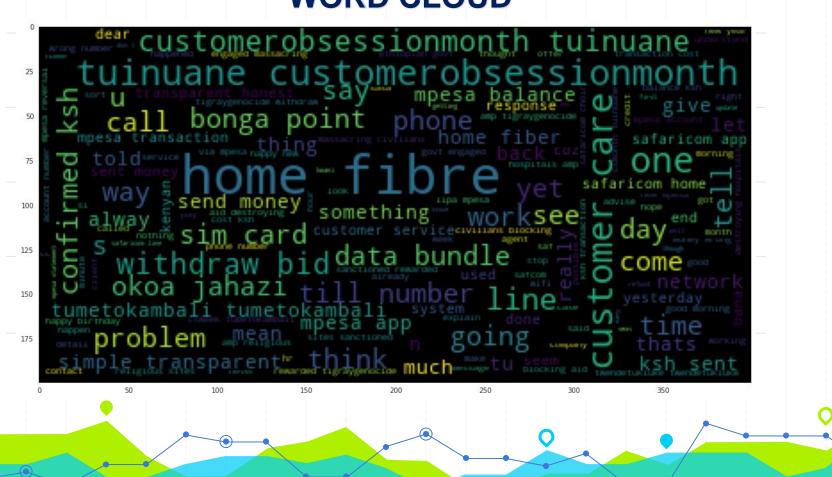






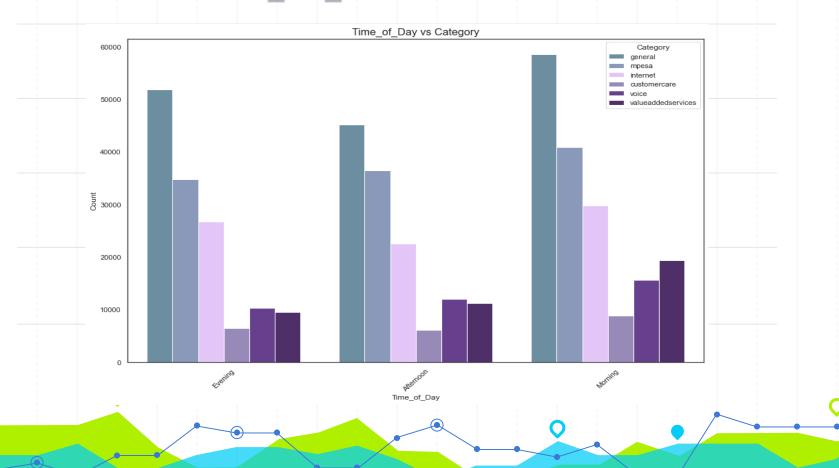


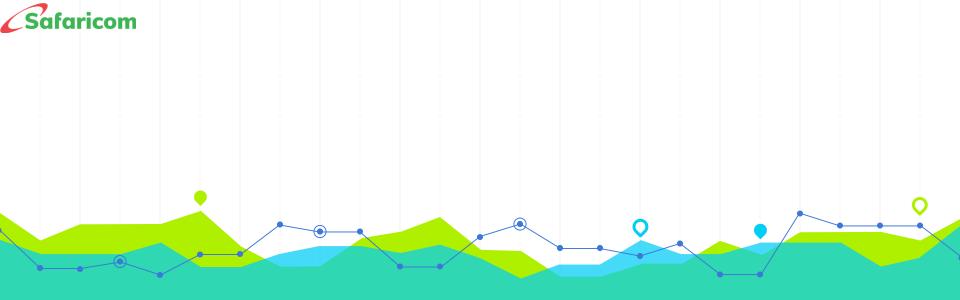
WORD CLOUD





TIME_OF_DAY VS CATEGORY





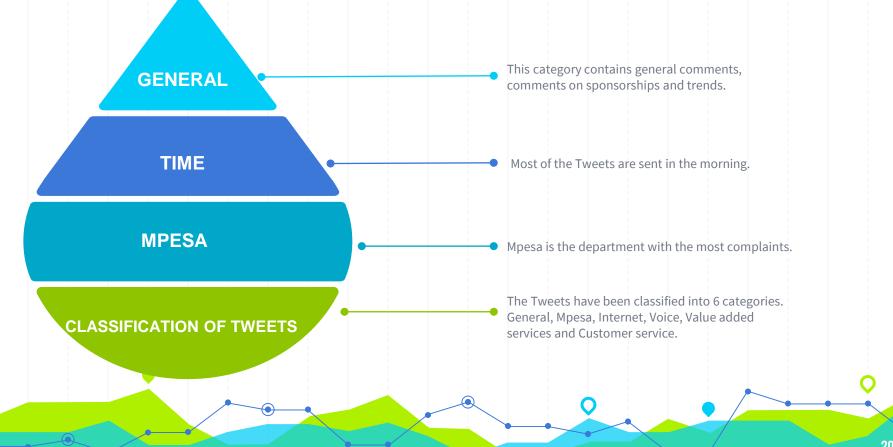
MODELING AND EVALUATION



MODEL EVALUATION

	TRAIN ACCURACY	TEST ACCURACY	RECALL	PRECISION	F1 SCORE
KNN	0.568	0.547	0.547	0.724	0.510
DECISION TREES	0.711	0.710	0.710	0.775	0.690
RANDOM FOREST	0.931	0.808	0.808	0.829	0.806
XGBOOST	0.947	0.947	0.947	0.949	0.947
NEURAL NETWORKS	0.999	0.999	0.999	0.999	0.999

CONCLUSION



RECOMMENDATION

Recommendations

Classify the tweets according to the respective departments

More attention should be given to the MPESA department.

Track time for planning employee shifts and strategizing marketing hours.

Look into the general category due to sponsorships and other general comments.

Benefits

Improved Customer Efficiency and better customer targeting in Marketing

Improved efficiency; New customers will increase their Revenue.

Reduce employee idle time, reduce cost and improved marketing strategy.

The Brand is maintained.

THANK YOU!

Any questions?