



DOM304: Contemporary Business Analytics

Group Project Presentation

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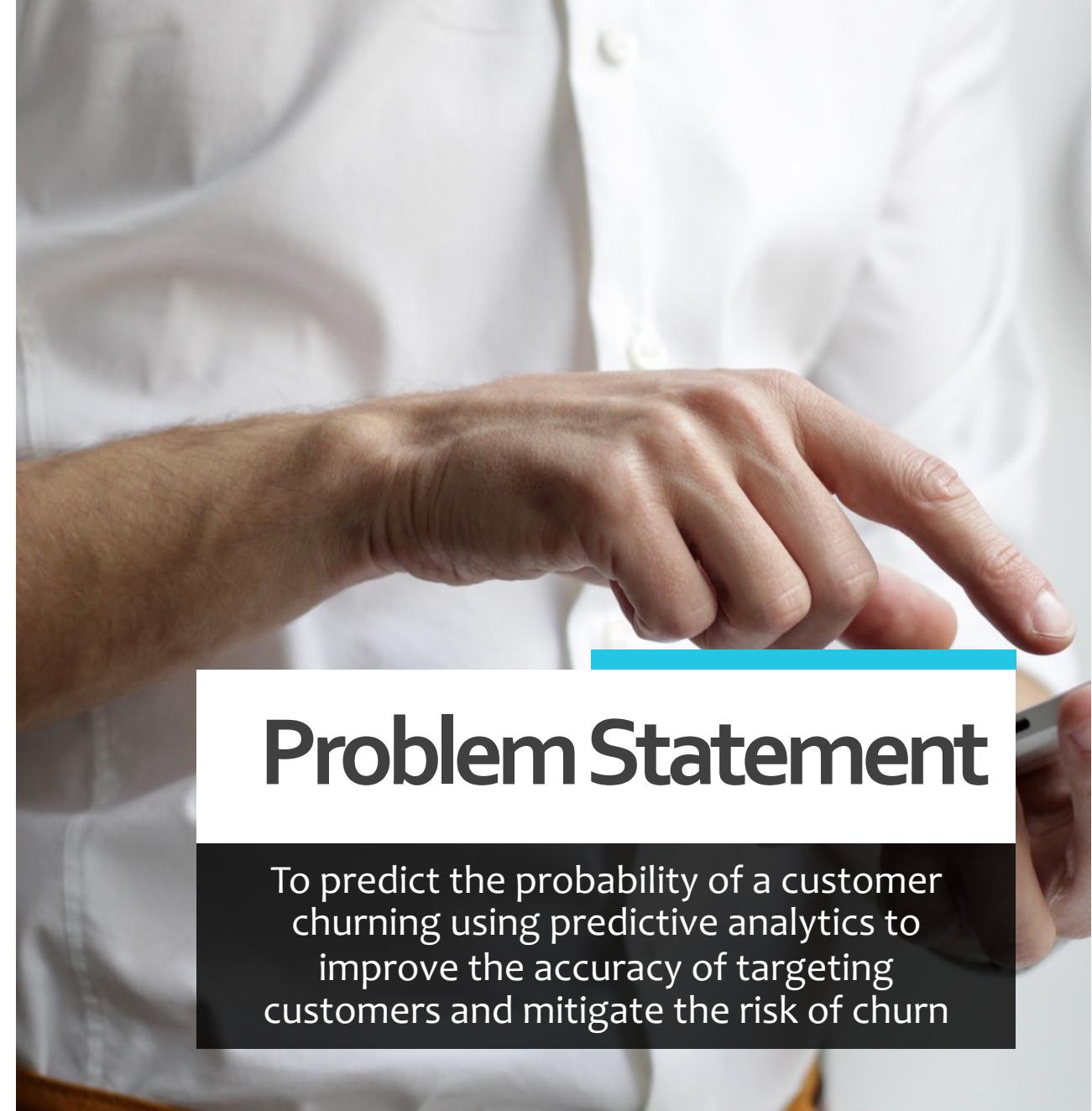
- Review of Models and comparing their results to benchmarks

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Business Implications

- Findings, Recommendations and Conclusions

- Cell-Tel is a leading Mobile Telecom Operator (MTO) with a stellar track record and a strong customer base. It offers a multitude of services to its 125 Mn customers across regions.
- With the recent liberalization of telecom policies by the government, Cell-Tel is under threat from newer MTOs. These MTOs have had relatively short tenures but have been successful in taking away market share from Cell-Tel.
- This has prompted the board to take corrective measures and the first in a series of measures is reducing the customer churn rate.
- The telecom operator has conducted a customer survey on a sample of the population of a region under threat from the newer MTOs.
- This survey data is used by the marketing team and retention managers to target customers who are at a high risk of churn and offer them incentives





Data Understanding

An overview of the attributes of the dataset and results of basic data exploration

Data

Features in the data set

- Generic features like gender, details of dependents, senior citizenship status, relationship status (partner yes or no)
- Add-ons include subscription to multiple lines, availing of internet services, subscription to streaming services both movies and TV
- Customer support level data points include availing of online backup service, seeking help from tech support, anti-virus subscription
- Revenue related features like contract type, billing mode, payment method, monthly charges and total charges

Data Quality

- Data quality was high with very little ambiguous values and outliers
- Mean imputation was employed to make up for missing values (<0.15% of the total data)
-

Overview

Brief insights into the data



7K
Customers
surveyed



32 weeks
Average tenure of
customers

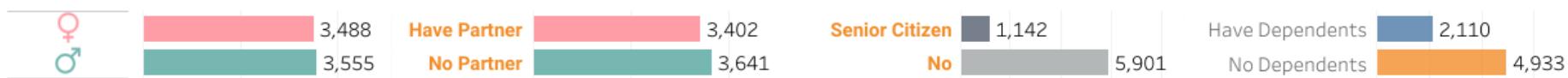
A graphic featuring a blue calendar with a red header and a blue clock with a yellow minute hand.

\$64.76
Average monthly
charges of customers

A graphic featuring a smartphone displaying a green checkmark inside a circle, accompanied by a gold coin and a green dollar bill.

Pictorial Representation of the data

Customer Demographics

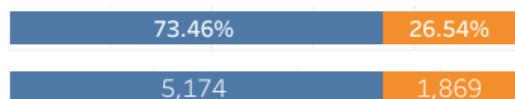


Managed Services Usage

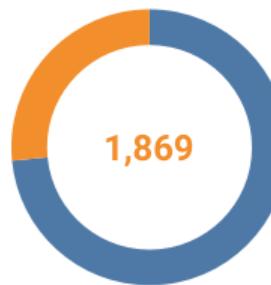


Churn Rate

Orange indicates the **customers lost**

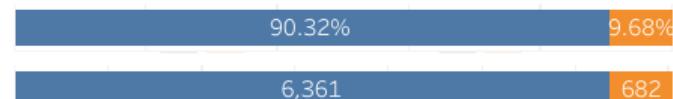


Total Customers : 7,043

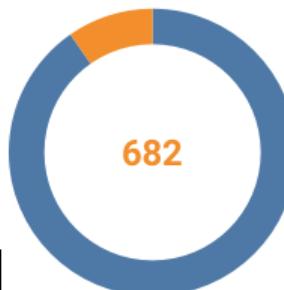


Phone Service Usage

Orange indicates the **customers not using service**



Project Data Source: Kaggle | <https://www.kaggle.com/blastchar/telco-customer-churn>





Modelling

A total of 4 models were tested

- Logistic Regression
- K Nearest Neighbours Classifier
- Naives Bayes Classifier
- Decision Tree Classifier

Model Metrics

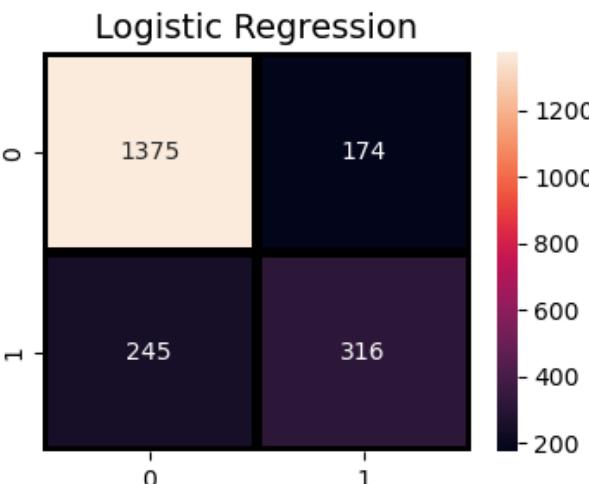
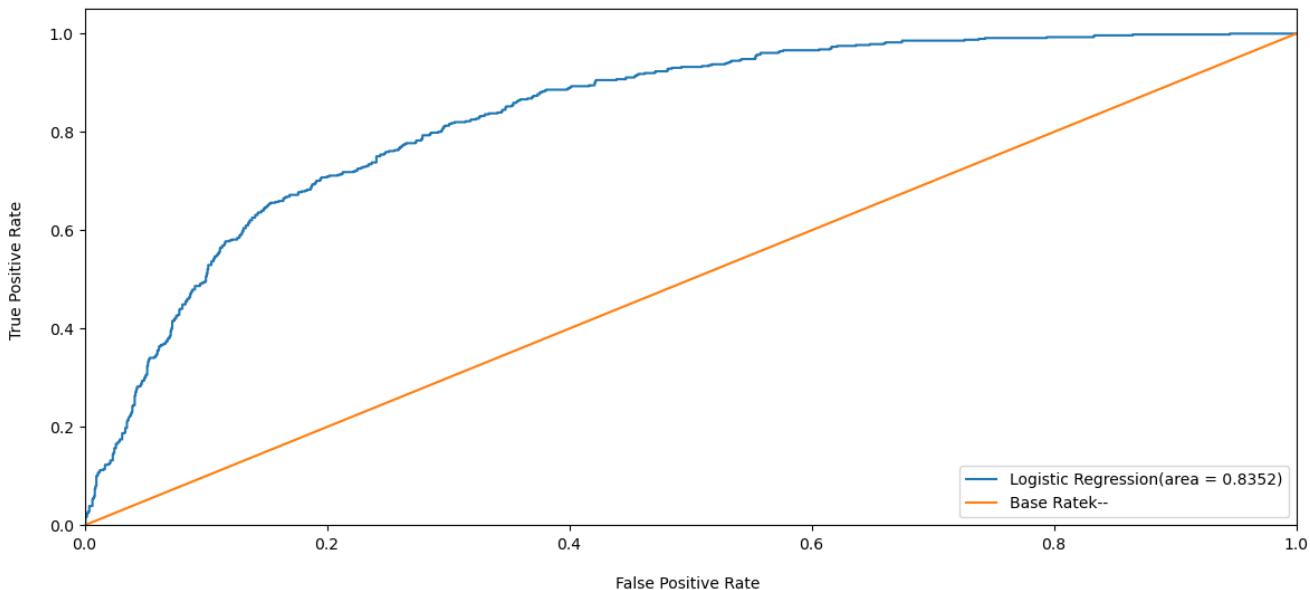
	Precision	Recall	F1 Score	Accuracy
Logistic Regression	0.644898	0.563280	0.601332	0.801422
K-Nearest Neighbours	0.579060	0.483066	0.526725	0.769194
Naive Bayes	0.508075	0.729055	0.598829	0.740284
Decision Tree	0.498239	0.504456	0.501329	0.733175

Logistic Regression

LR is chosen and is recommended for deployment due to its appropriateness to the use case

- Logistic Regression is relatively easy to implement and explain
- The algorithm works very well on unseen data – an important parameter for Cell-Tel which operates in a sector where customer preferences change quite frequently
- The algorithm is less prone to over-fitting, thereby aiding the use case of the dynamic customer behaviour
- Since this is a sample of the population, Cell-Tel can very well scale this model to its large customer base of 125 Mn without compromising on computational efficiency

ROC Graph



Accuracy
80.142%

F Score
0.6013

Conclusion and Recommendations

- Comments on business implication
- Comments on shelf-life on model
- Guidance to reduce churn rates



Business implication and model shelf life

- Month-to-month contract type users are at the highest risk of churn
 - Converting current month-to-month users to long term contract subscribers has statistically shown to reduce the customer churn rate. Recommend amplifying the conversion rate by providing content about the key functional benefits of applying for a subscription model
- More than half of the churned customers have not subscribed to additional internet services and/or streaming services
 - Recommend attractively introducing additional internet and streaming services to these users. There is a higher chance that they will retain and become high value customers after purchasing these add-on offers.
- Fiber optic internet service is popular among both (churned and retained) customer groups
 - Cell-Tel can develop a better pricing plan since majority customers prefer unlimited data and revenue from fiber optic internet service could be used to offset any loss incurred due to customer churns.
- The model is expected to perform well unless there is a tectonic shift in consumer behaviour or there are newer government regulations.
- Cell-Tel is recommended to conduct periodic customer surveys to not only improve the current model, but also to leverage data to improve:
 - Ticket classification
 - Anomaly detection
 - Predictive maintenance
 - Dynamic Optimization



Thank You

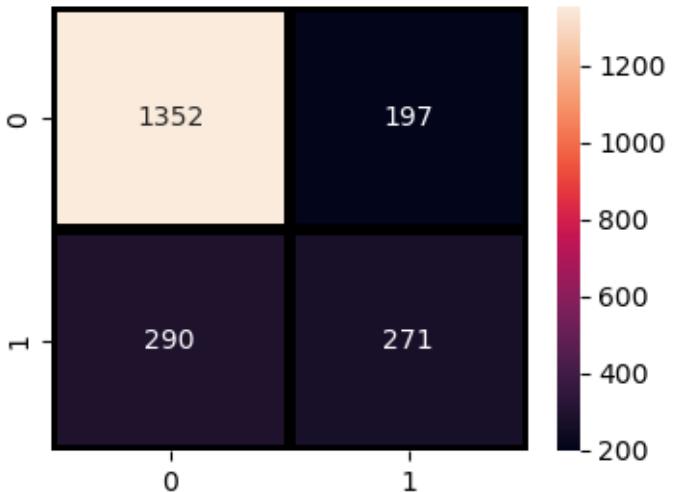
Appendix



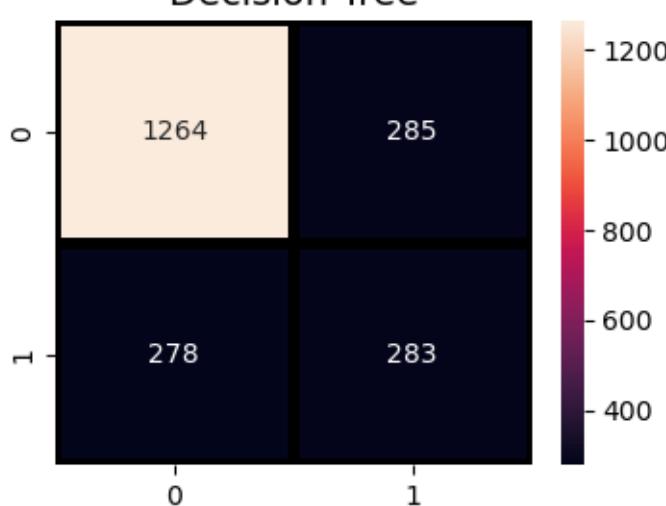
Other Model Performances

Confusion Matrices

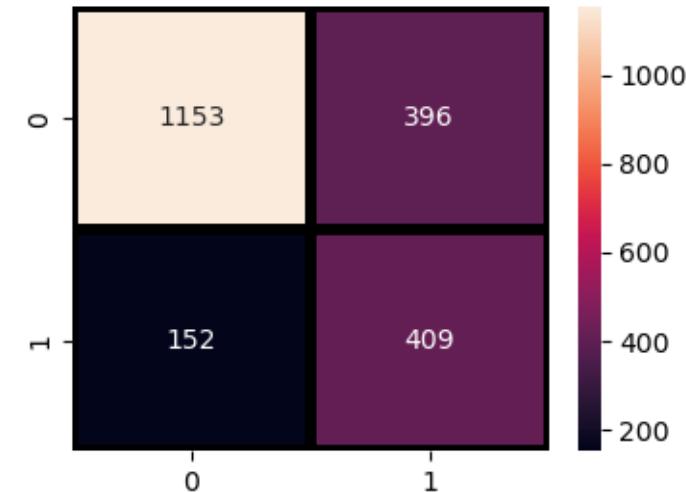
K-Nearest Neighbours



Decision Tree

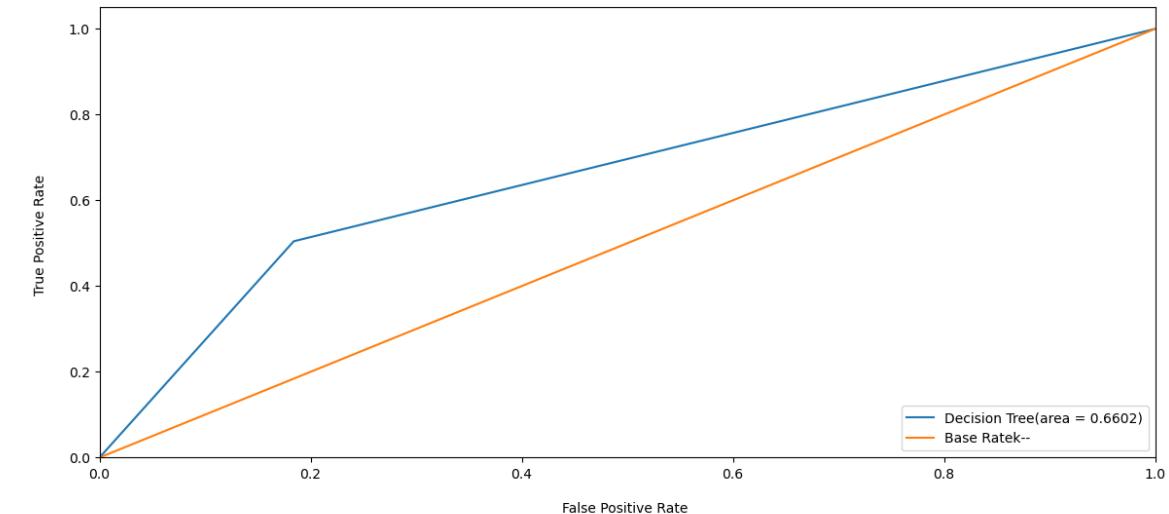


Naive Bayes

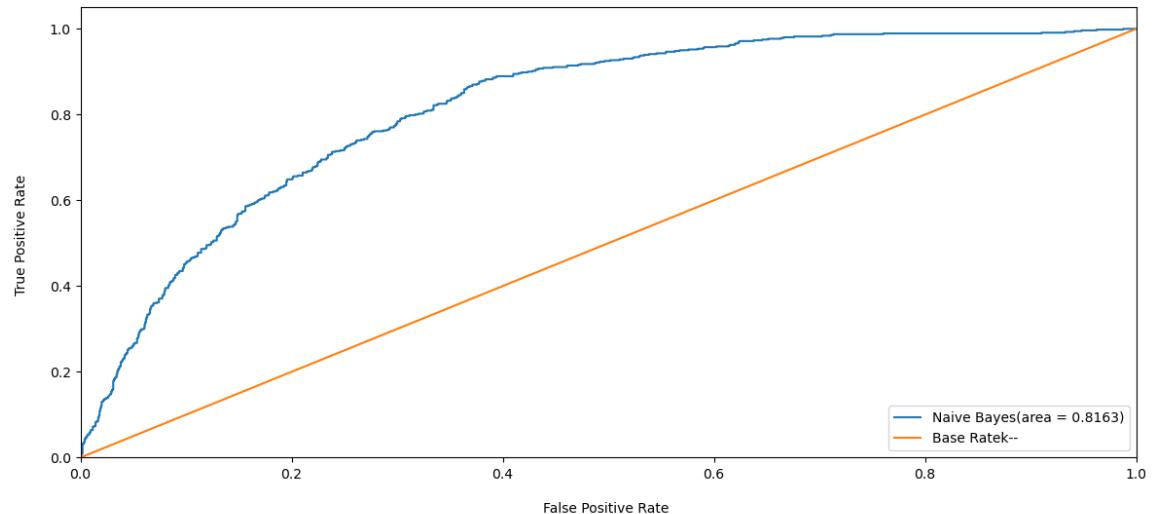


Receiver Operating Characteristic (ROC) Curve

ROC Graph



ROC Graph



ROC Graph

