



CUSTOMER CHURN PREDICTION OF TELECOM COMPANY

SASSY PREDICTORS
GROUP 8

CONTENTS



Introduction

3



Data

8



Analysis

16



Results

20



Conclusion

26

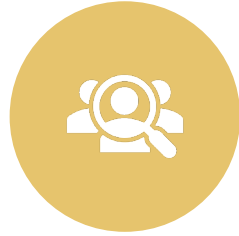


SECTION I: INTRODUCTION

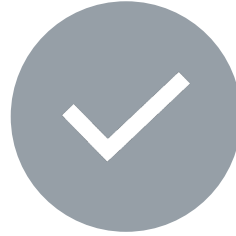
LAYING FOUNDATION OF OUR PROJECT'S VISION, BRIEF
INDUSTRY BACKGROUND AND THE GOALS WE AIM TO MEET
VIA ANALYSIS



Customer loyalty is the key to profitability in the telecom industry



Analysis to focus on the behavior of telecom customers who are more likely to leave the platform (i.e., Customer churn).



Intend to initially perform EDA to gauge customer behavior



Thereafter, use predictive analytics techniques to determine the customers who are most likely to churn



Lastly, analyze model and results to predict behavior for a customized customer retention programs, which in turn would help in increasing profits

INTRODUCTION



Telecom services such as voice call and internet are a necessity these days.



Telecom Market has become saturated over the years lacking differentiating factors between competing brands.



Acquiring new customers can be as much as 5X expensive than retaining existing ones thus, companies are constantly looking for innovations to retain customers.

TELECOM INDUSTRY BACKGROUND

WHY UNDERSTAND CUSTOMER CHURN?

Customer acquisition is expensive thus, it is important to retain existing customers.

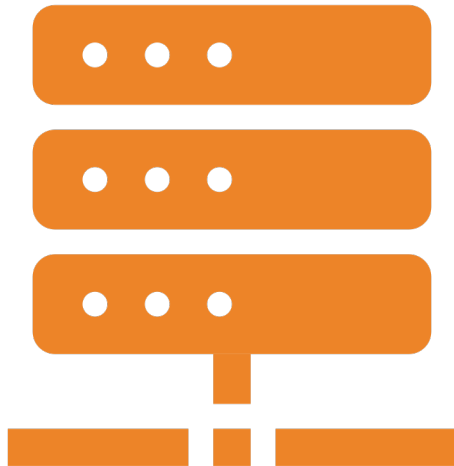
Is all churn bad? Understanding customer lifetime value is important for that.

Churn is usually result of combination of several customer issues.

OBJECTIVES

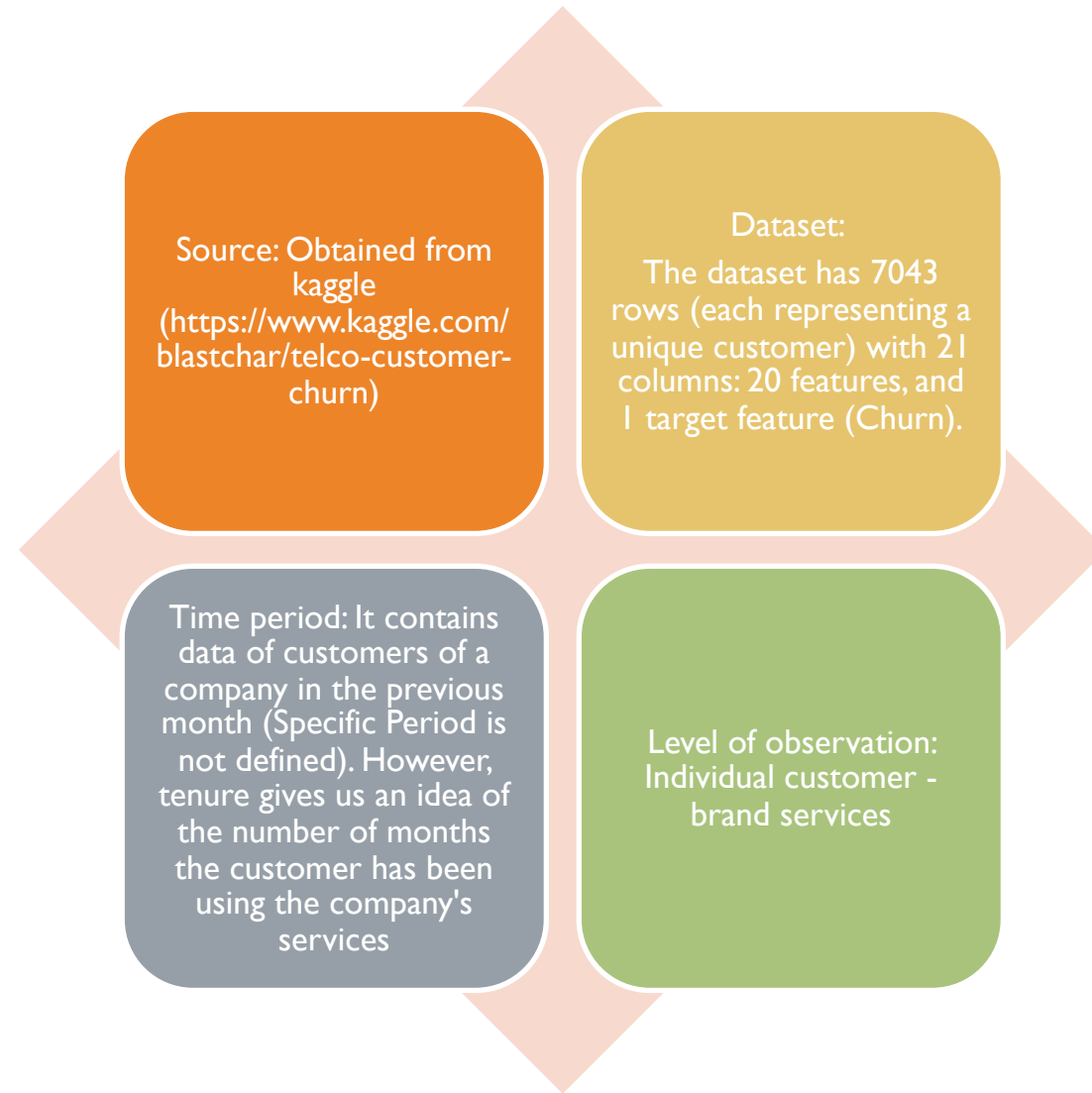


- Identify potential factors leading to customer churn
- Understanding demographic information affecting churn
- Analyzing behavior based on the type of contracts and payment methods
- Identifying loyal customers and rewarding them
- Engaging with customers likely to churn
- Develop marketing and promotional strategies



SECTION 2: DATA

DESCRIBING THE VARIABLES AND PARAMETERS IN OUR DATA AND
DEMOGRAPHICS DATA EXPLORATION AND VISUALIZATION



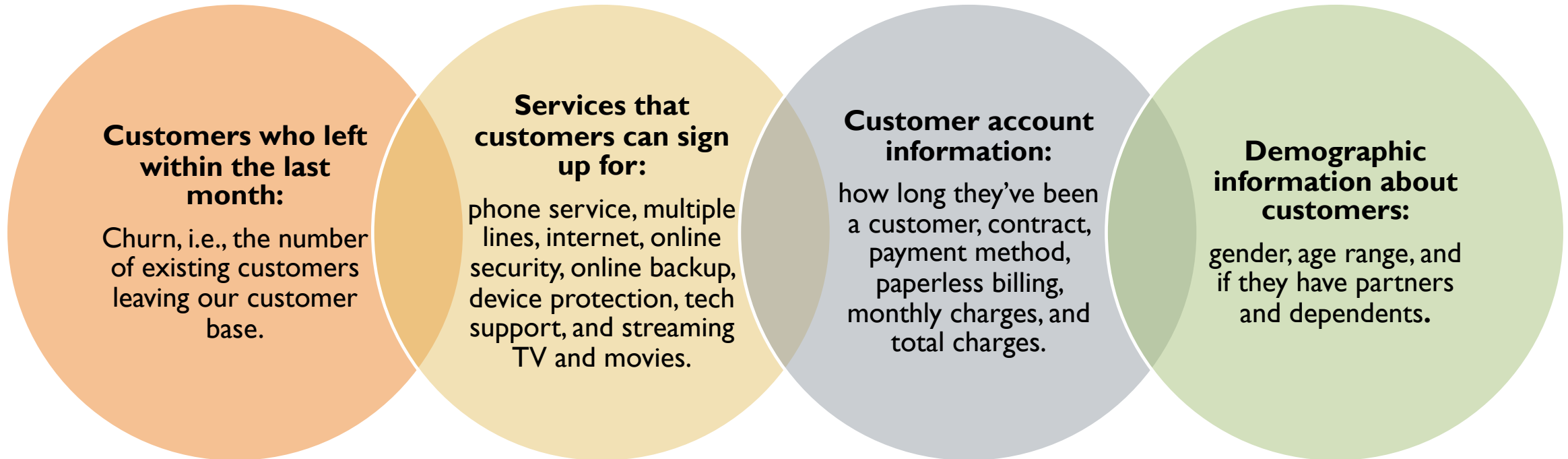
DATA OVERVIEW

Attribute	Description	Type
Customer ID	Each unique identification number of a customer	VARCHAR
Gender	Gender of the customer (Female / Male)	CHAR
Senior Citizen	Age Range to identify customers' age group (0/1)	BINARY
Partner	Whether the customer has a partner (spouse or partner) or NOT (Yes/No)	BOOLEAN
Dependents	Whether the customer has a dependent (for e.g.,: child, spouse, parent, or certain other relative to whom they provide financial support) or NOT (Yes/ No)	BOOLEAN
Tenure	Months for which subscriber has been with company	NUM
Phone Service	Whether the Customer has a phone service or NOT (Yes/No)	BOOLEAN
Multiple Lines	Whether customer has subscribed to multiple lines or NOT (Yes / No / No phone service)	CHAR
Internet Service	Whether the customer has an internet service, if yes - which data service (DSL / Fiber optic / No)	CHAR
Online Security	Whether the customer has online security or NOT (Yes / No / No internet service)	CHAR
Online Backup	Whether the customer has online backup or NOT (Yes / No / No internet service)	CHAR
Device Protection	Whether the customer has device protection or NOT (Yes / No / No internet service)	CHAR
Tech Support	Whether the customer has technological assistance or NOT (Yes / No / No internet service)	CHAR
Streaming TV	Whether the customer has television streaming or NOT (Yes / No / No internet service)	CHAR
Streaming Movies	Whether the customer has movie streaming or NOT (Yes / No / No internet service)	CHAR
Contract	Type of customer's contract term (Month-to-month / One year / Two year)	CHAR
Paperless Billing	Whether the customer has paperless billing or NOT (Yes / No)	BOOLEAN
Payment Method	The type of payment method the customer uses (Electronic check / Mailed check / Bank transfer (automatic) / Credit card (automatic))	CHAR
Monthly Charges	The amount of charges the customer pays monthly	NUM
Total Charges	The total amount of charges the customer pays	NUM
Churn	Whether the customer has left the provider or NOT (Yes / No)	BOOLEAN

DATA DICTIONARY



KEY VARIABLES

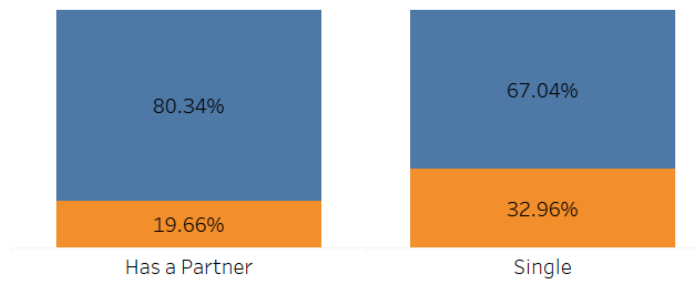


CONSUMER DEMOGRAPHICS

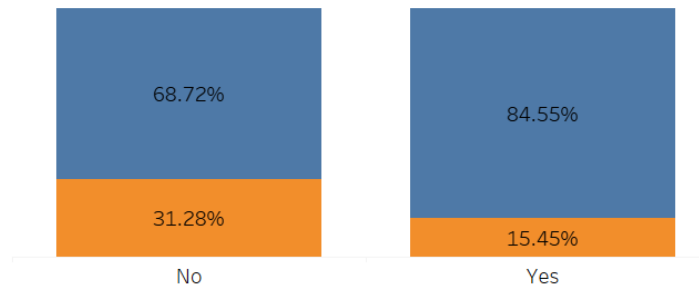
- We see that large portion of customers who churn have no partners or dependents. (Focusing on better, cheaper family plans can be useful to reduce churn)
- We see considerable amount of churn who have monthly contracts around 23%. (this is because most monthly customers are confused between carriers and don't want to commit to anyone)
- Gender seems to have no significant difference on why a customer leaves the company

CUSTOMER CHURN ANALYSIS

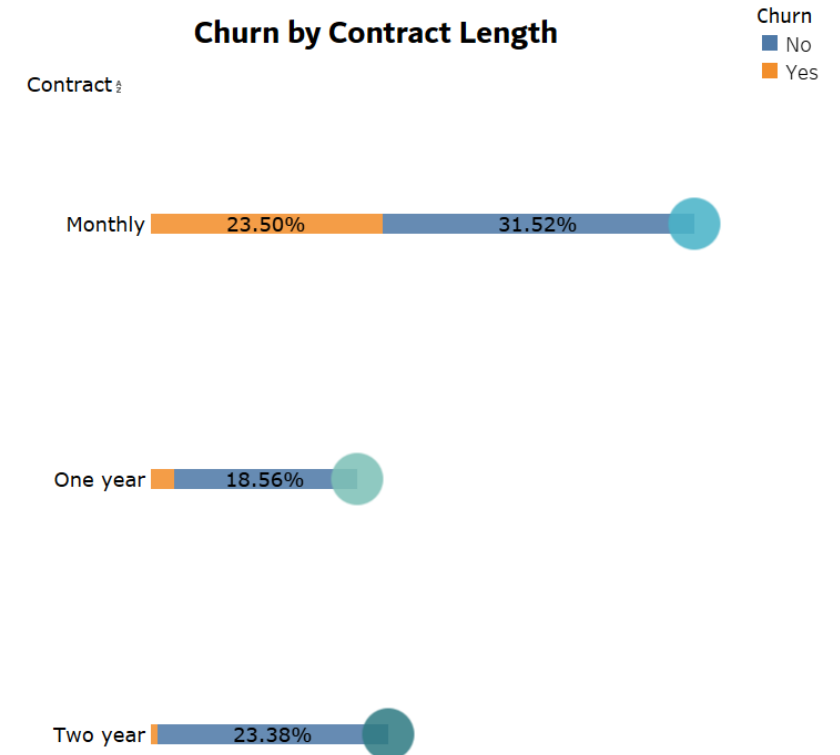
% of Churn by Partner Status

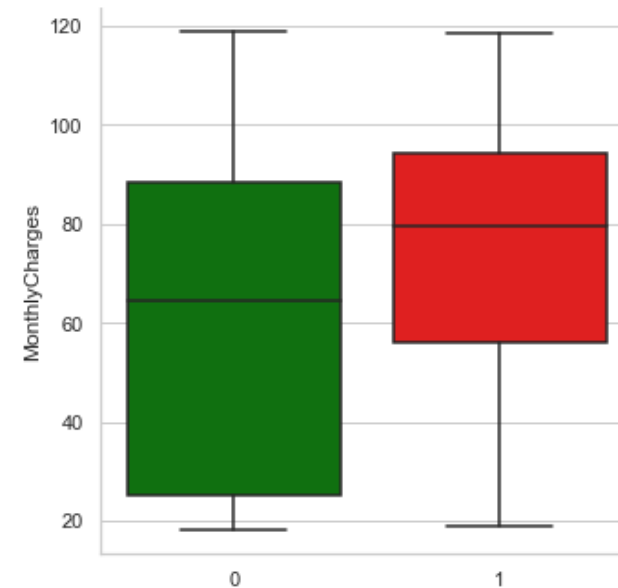
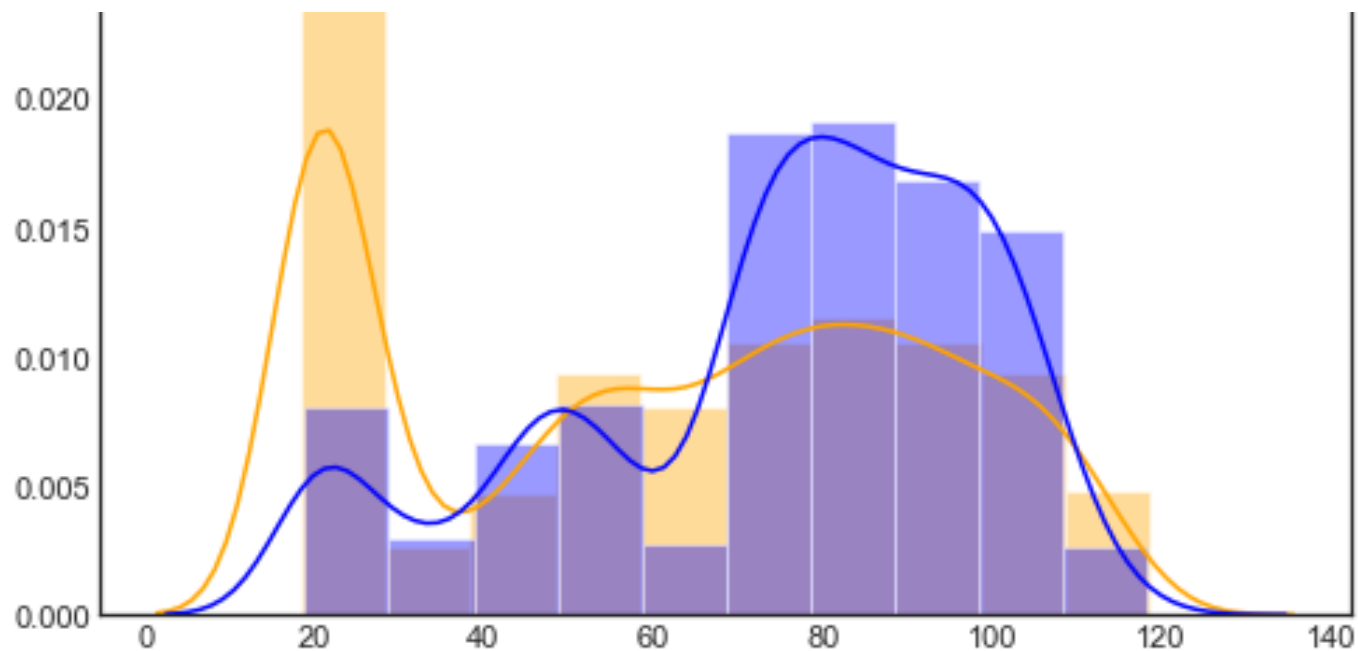


% of Churn by Dependents



Churn by Contract Length



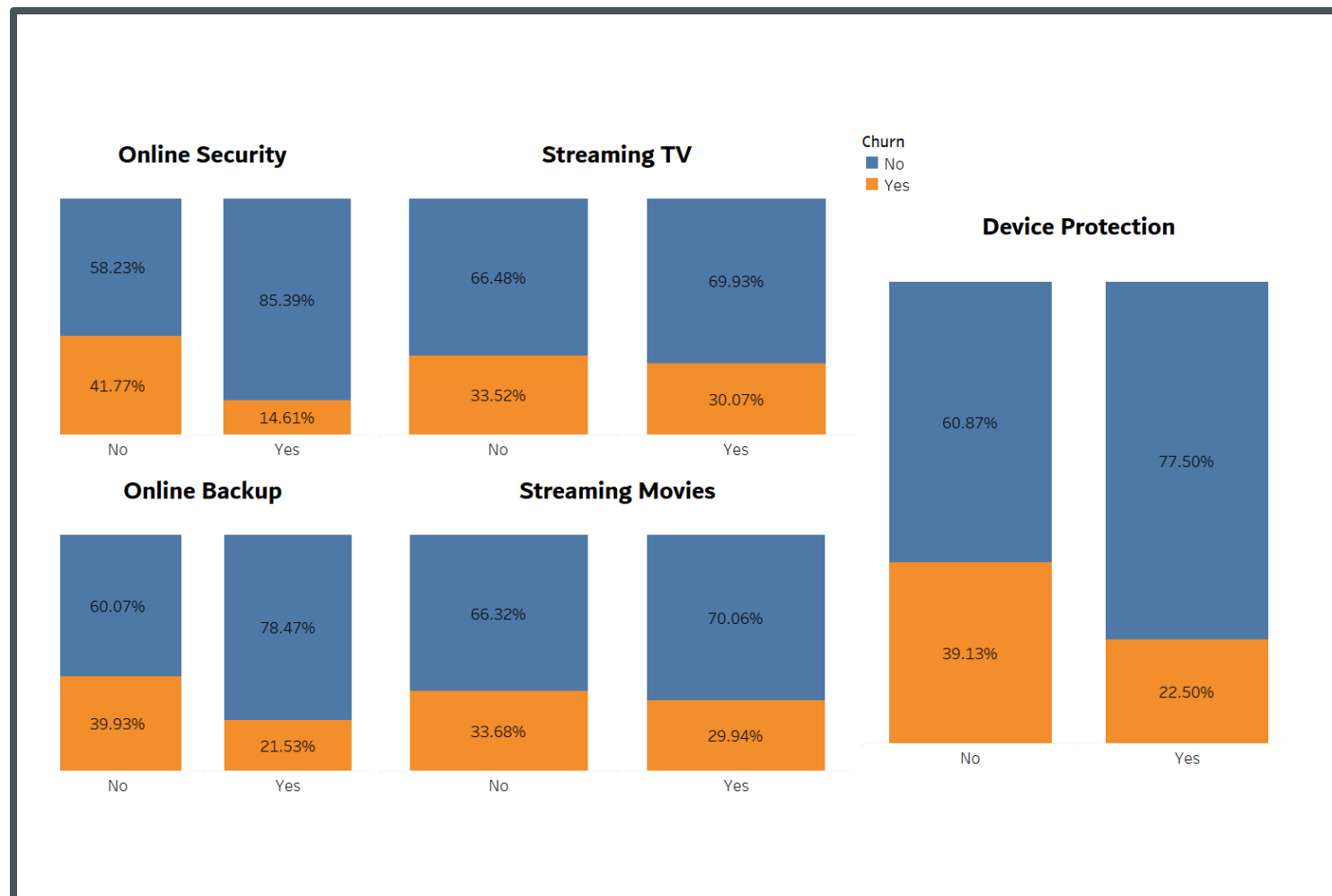


MONTHLY CHARGES & CHURN

We see the maximum churn amongst our customers at monthly charges around \$70-110.

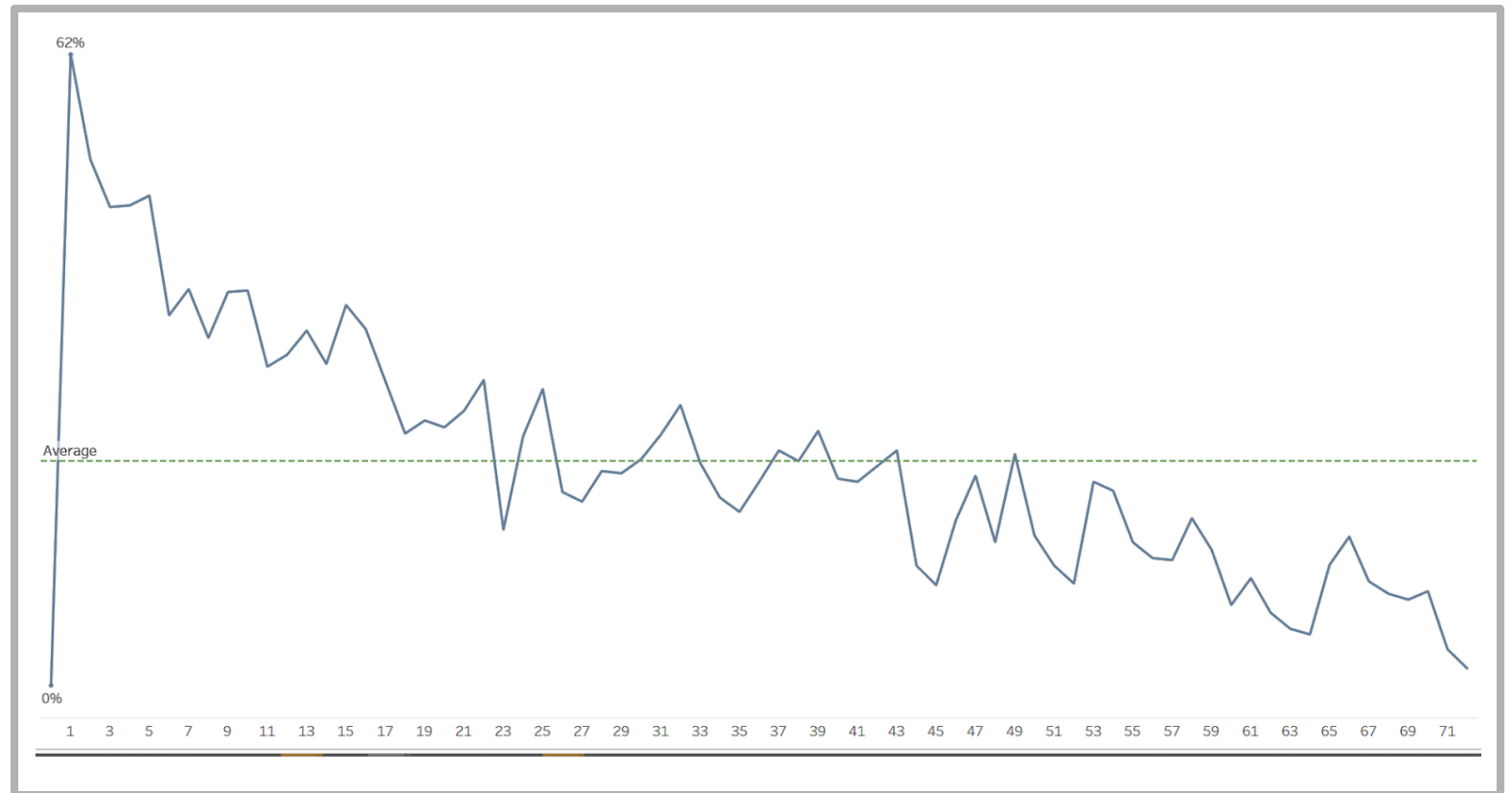
SERVICES USAGE

- *With increasing number of services we see a considerable drop in the churn rate for customers.*
- *This is more evident in services where there is some barrier to change to other platform unlike Streaming services.*

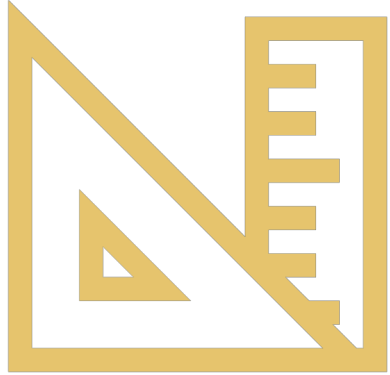




AT WHAT POINT
IN TENURE IS THE
CHURN HIGHEST?



- *Customers who are with the telecom providers for a shorter duration (usually prepaid customers) usually switch to another provider without a prior notice.*
- *Predicting their churn is tricky as one might wonder if the services are being used on and off (for a short travel every few months), or the customer has churned.*



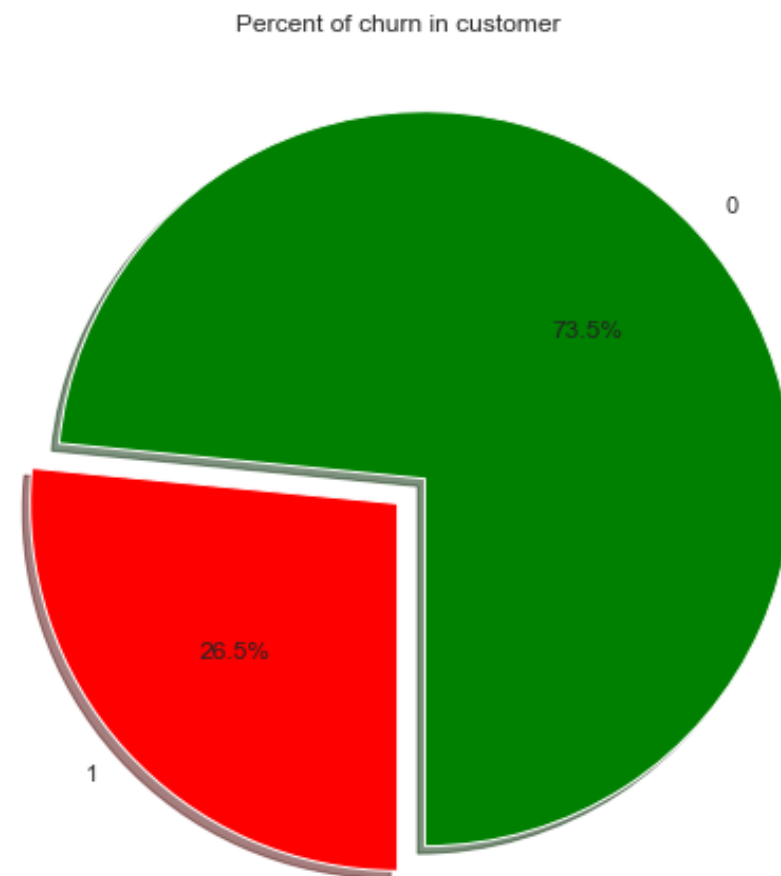
SECTION 3: ANALYSIS

MODEL SELECTION, METHODOLOGY AND DETAILED ANALYSIS

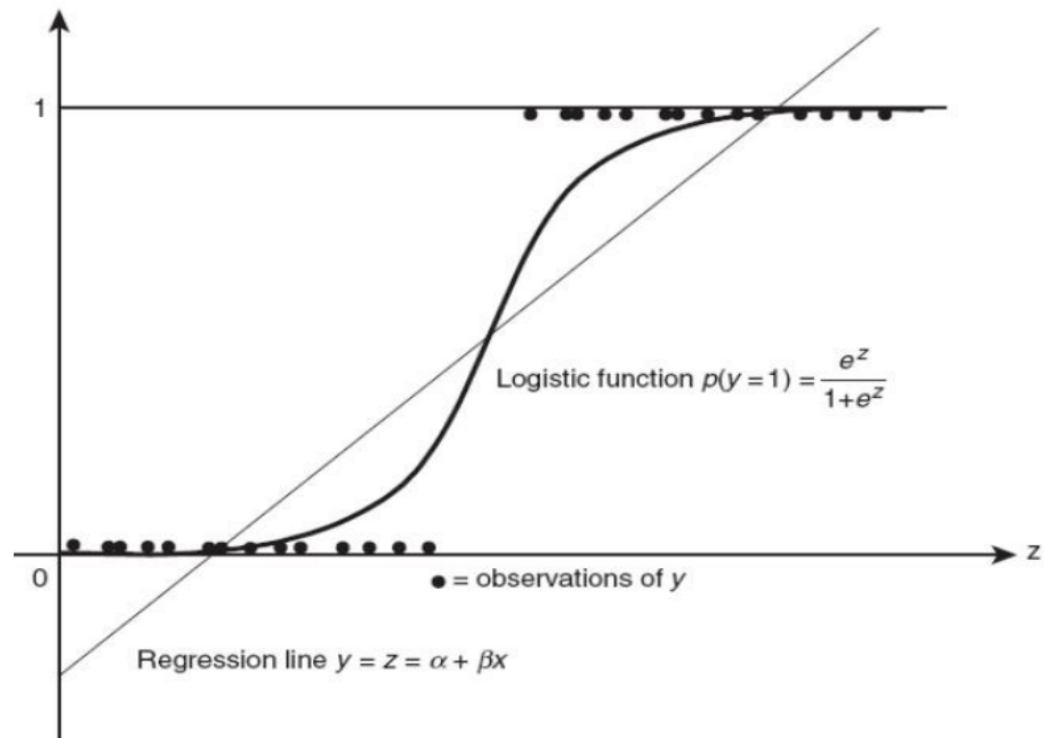
MODEL SELECTION

Customer Churn
Classification
(categorical: yes /
no)

Choice of Model:
Logistic
Regression



WHY LOGISTIC REGRESSION



Logit function vs
Linear Probability
Model

Probability values
between
- $[0, 1]$
Homoscedasticity -
residuals not normal



Odds Ratio : Better
interpretation

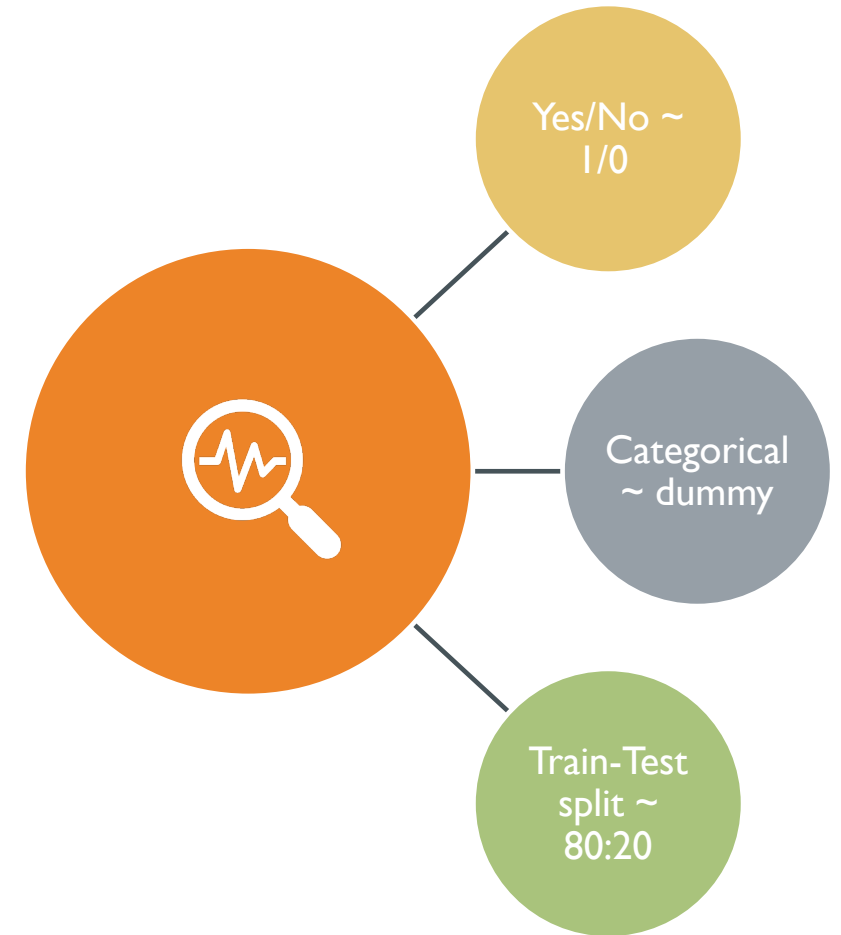
FEATURE SELECTION & DATA PREPARATION

Based on exploratory analysis and statistical significance, following features are **not considered**:

Gender, Senior Citizen, Online Security, Online Backup, Device Protection, and Tech Support

Monthly Charges has been selected over Total Charges to address **multi-collinearity**.

Interaction effects included: Fiber Optic*Monthly Charges, Streaming Movies*Streaming TV, and Dependents*Partners

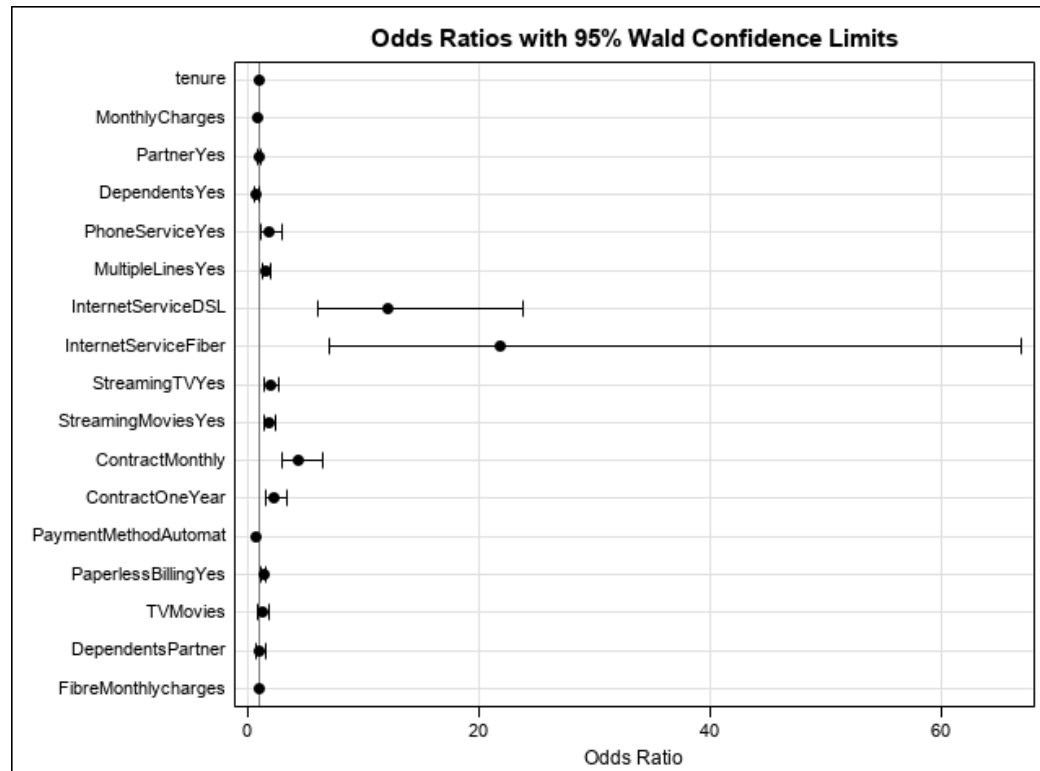




SECTION 4: RESULTS

FINDINGS OF THE ANALYSIS, OUR CONCLUSIONS, MACRO IMPACT AND
RECOMMENDATIONS

LOGISTIC REGRESSION MODEL



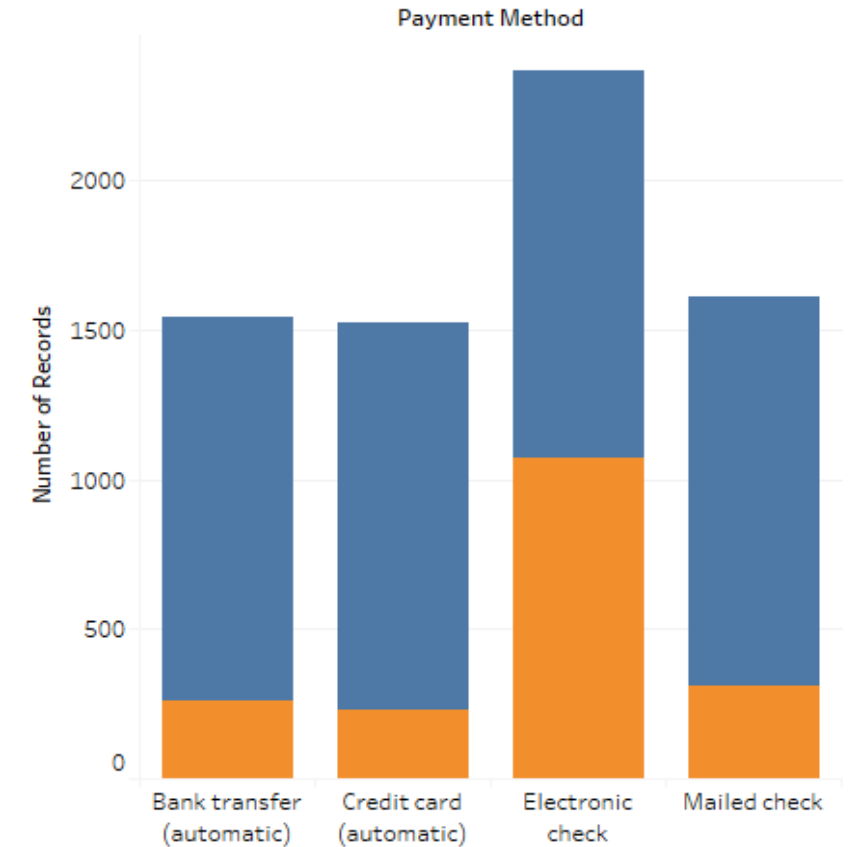
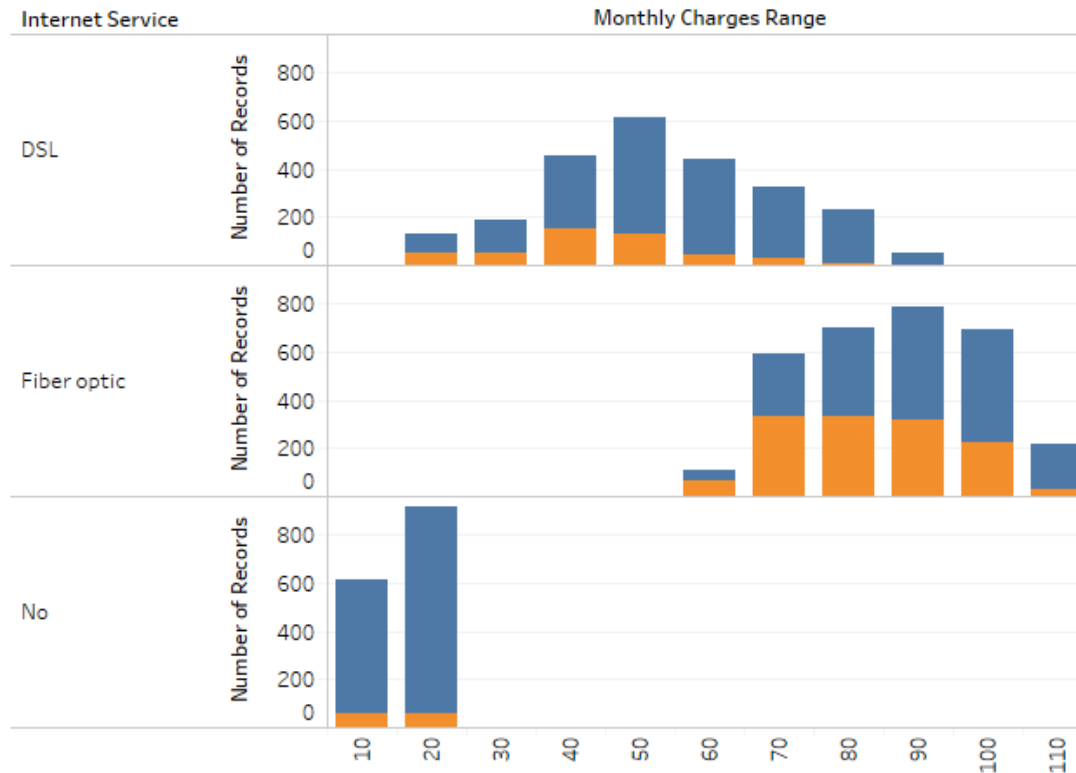
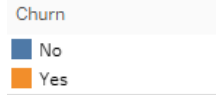
Odds Ratio: variables having higher odds ratio and statistical significance:

- Internet Service with Fiber Optic
- Internet Service with DSL
- Monthly Contracts
- One Yearly Contracts

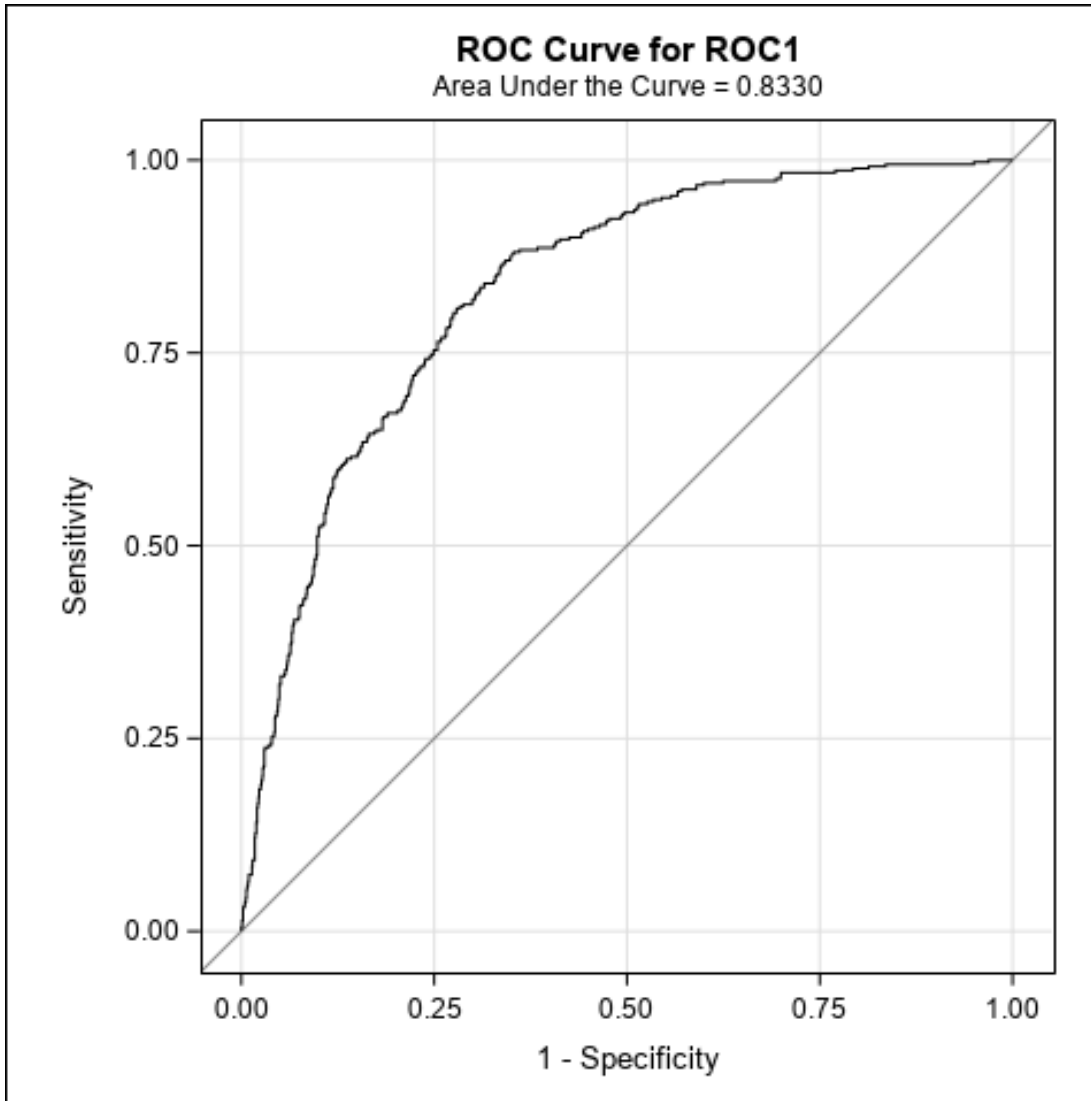


Variable having lower odds ratio and statistical significance:

- Payment Method Automatic



EFFECTS OF MONTHLY CHARGES & INTERNET SERVICE, AND PAYMENT METHOD ON CHURN AS OBSERVED IN THE LOGISTIC MODEL



ROC CURVE
AUC = 0.8330



APPLICATION BASED THRESHOLD SELECTION

Promotional
offers &
Bundle
provision

Cost of predicting false
churn (False Positive) -
\$20

Cost of not
predicting true churn
(False Negative) - \$80

Minimum Cost = \$10,940

Threshold Probability =
0.201



APPLICATION BASED THRESHOLD SELECTION

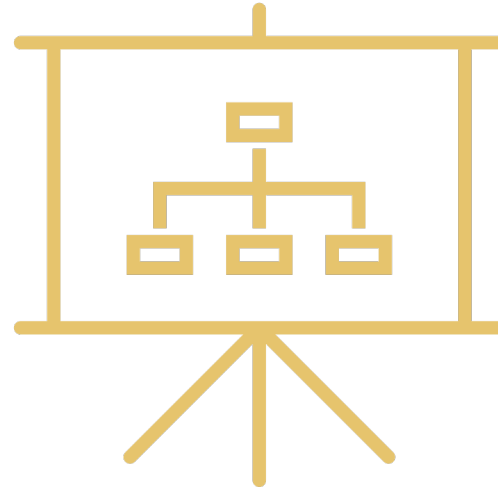
Engaging
with
Customers
by call /
email:

Cost of predicting false
churn (False Positive) - \$3

Cost of not
predicting true churn
(False Negative) - \$10

Minimum Cost = \$1,541

Threshold Probability
= 0.203



SECTION 5: CONCLUSION

DESCRIBING THE VARIABLES
AND PARAMETERS IN OUR
DATA AND DEMOGRAPHICS
DATA EXPLORATION AND
VISUALIZATION

1

Optical fiber users were most likely to churn in the previous month.

2

Customer churn is very high among monthly contract users.

3

Autopay enrollment is the best performing feature for retention.

4

The belief that multiple lines using customers are less likely to churn is wrong.

CONCLUSION

RECOMMENDATIONS



Immediate actions
required for internet
prices, especially for
optical fibre.



Try to convert
monthly users to long
term plans.



Encourage customers
for automatic payment
enrollment by giving
discounts/credits.



Reward customers -
least likely to churn
and long tenure.



Service bundling for
reward/promotion to
attract for
longer terms



To engage with
customers who are
likely to churn.



THANK YOU!

WE ARE OPEN FOR QUESTIONS



Arunachalam Ramanathan

Mihir Bansal

Mitansh Patel

Priyanka Agarwal

Rohini Kanungo

Sagar Patel