

Customer Churn Analysis

using Power BI:

Internship Report



Intern: Kshetrimayum Galax Singh

Ref : SKS/A2/C0106

Company: Saiket Systems (<https://www.saiket.in/>)

Domain: Machine Learning

Batch: Jan-Feb 2024

GitHub repo: <https://github.com/galax19ksh/Customer-churn-analysis/>

Contact: +91 7005788363

Mail: galaxkshetrimayum16@gmail.com

What is Customer Churn?

Customer churn, also known as customer attrition, refers to the loss of customers over time. It's basically when customers stop using your product or service and decide not to return. It's important for businesses to track and understand their churn rate, as it can have a significant impact on their bottom line.



Telecom industry:

Customers in the telecom industry can choose from a variety of service providers and actively switch from one to the next. The telecommunications business has an annual churn rate of 15-25 percent in this highly competitive market.

Causes of churn:

- **Poor customer service:** Long wait times, unhelpful interactions, unresolved issues.
- **Product dissatisfaction:** Product not meeting expectations, lack of features, bugs.
- **Competitor offerings:** Better options available at competitive prices.

- **Pricing issues:** Prices perceived as too high, hidden fees, unexpected charges.
- **Changes in customer needs:** Needs evolve, your offering doesn't adapt.

Customer retention:

Keeping your customers happy while managing a large crowd can feel like juggling flaming bowling pins. Time and resources are limited, making it impossible to shower everyone with personalized attention. But what if you could predict who's about to jump ship?

Here's where customer churn prediction comes in. Just like a weather forecast, it uses data to tell you which customers are at high risk of leaving. This lets you focus your efforts on saving the "at-risk" ones, maximizing your impact and saving precious resources.

Think of it like this: Instead of sending everyone a generic umbrella, you only give them to those caught in a downpour.

Objectives:

- Finding the % of Churn Customers and customers that keep in with the active services.
- Analysing the data in terms of various features responsible for customer Churn
- Finding a most suited machine learning model for correct classification of Churn and non-churn customers.

Data Source: https://drive.google.com/file/d/1wSOU_MYqbDanEf_BkypegNKkxzOfbJXU/view?usp=drive_link

Tools used:

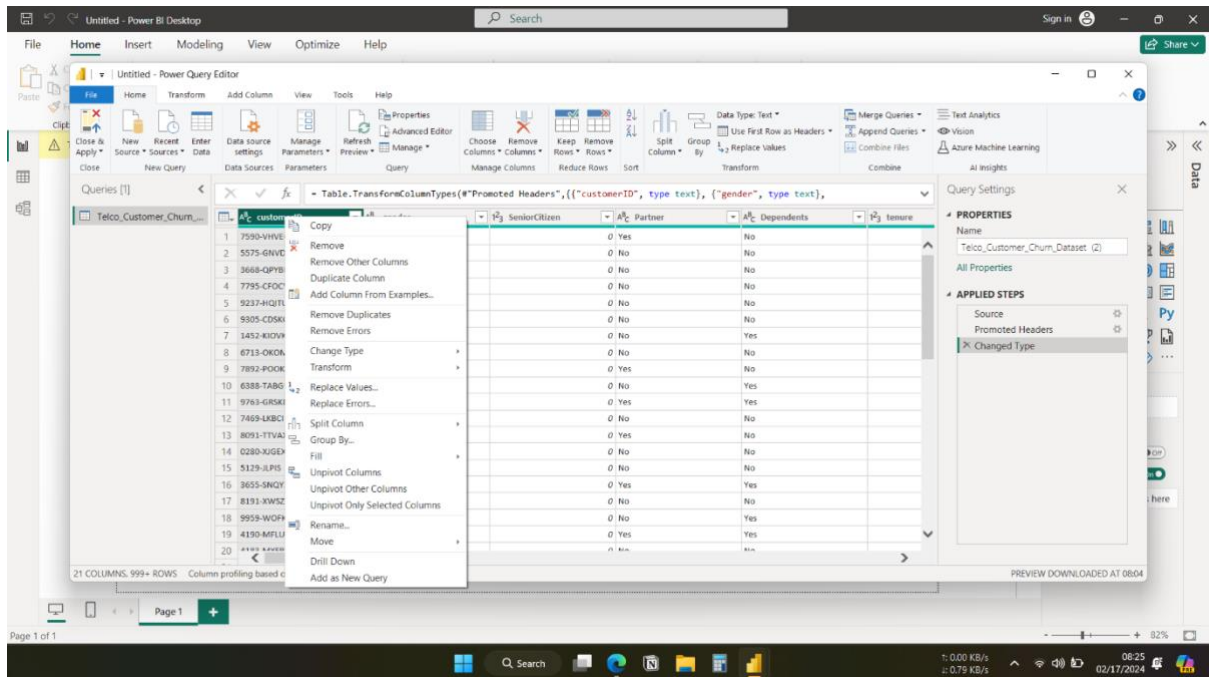
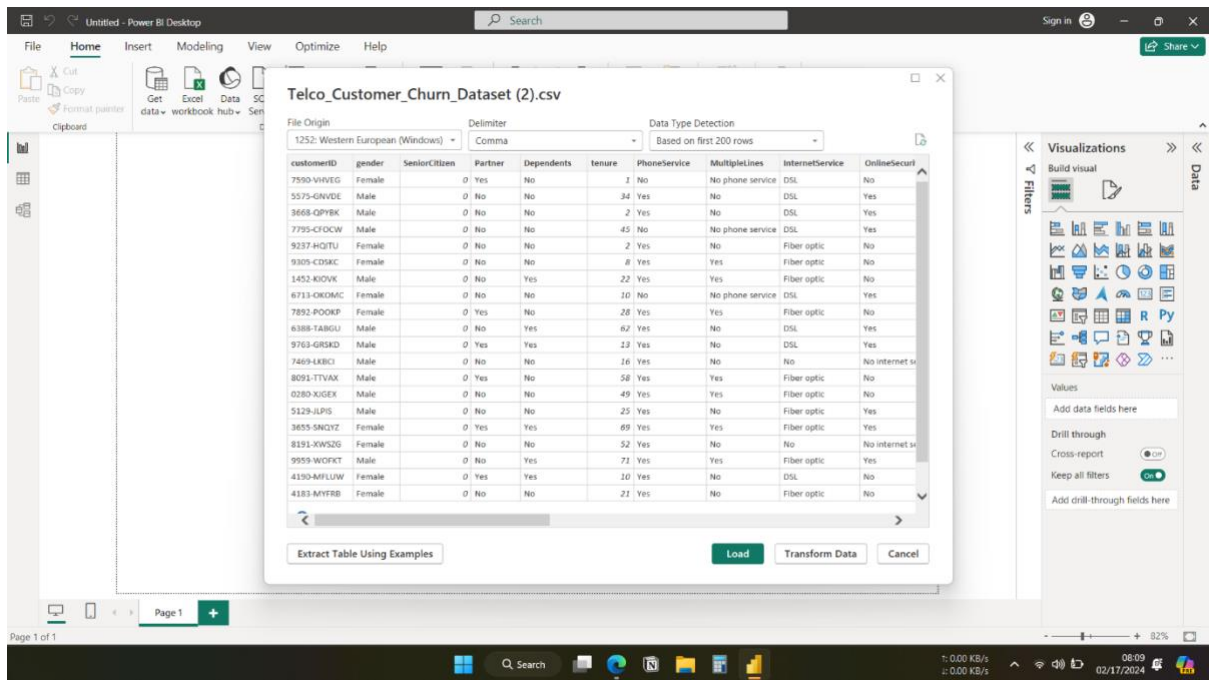
- Power BI
- Google Colab

Libraries used:

Pandas, numpy, matplotlib, sklearn, seaborn

Visualization using Power BI:

- Load CSV file into Power BI desktop
- Check if there is any missing values
- Remove unnecessary columns



Untitled - Power Query Editor

File Home Transform Add Column View Tools Help

Query Settings

Layout Data Preview Columns Parameters Advanced Dependencies

Queries [1]

Table.RemoveColumns(#"Changed Type",{"customerID"})

gender SeniorCitizen Partner Dependents tenure PhoneService MultipleLine

Valid 100% Valid 100% Valid 100% Valid 100% Valid 100% Valid 100% Valid 100%

Error 0% Error 0% Error 0% Error 0% Error 0% Error 0% Error 0%

Empty 0% Empty 0% Empty 0% Empty 0% Empty 0% Empty 0%

1 Female 0 Yes No No 34 No No phone ser

2 Male 0 No No No 2 Yes No

3 Male 0 No No No 45 No No phone ser

4 Male 0 No No No 2 Yes No

5 Female 0 No No No 8 Yes Yes

6 Female 0 No No No 22 Yes Yes

7 Male 0 No No No 10 No No phone ser

8 Female 0 Yes No No 28 Yes Yes

9 Female 0 No Yes Yes 62 Yes No

10 Male 0 Yes Yes Yes 13 Yes No

11 Male 0 No No No 16 Yes No

12 Male 0 Yes No No 58 Yes Yes

13 Male 0 No No No 49 Yes Yes

14 Male 0 No No No 25 Yes No

15 Male 0 Yes Yes Yes 69 Yes Yes

16 Female 0 No No No 52 Yes No

17 Female 0 No Yes Yes 71 Yes Yes

18 Male 0 Yes Yes Yes 10 Yes No

19 Female 0 No No No 21 Yes No

20 Female 1 No No No 1 No No phone ser

21 Male 0 Yes No No 12 Yes No

22 Male 0 No No No 1 Yes No

23 Male 0 No No No 58 Yes Yes

24 Female 0 Yes No No 58 Yes Yes

25

PROPERTIES

Name

Telco_Customer_Churn_Dataset (2)

APPLIED STEPS

Source

Promoted Headers

Changed Type

Removed Columns

20 COLUMNS, 999+ ROWS Column profiling based on top 1000 rows

PREVIEW DOWNLOADED AT 08:04

Untitled - Power BI Desktop

File Home Insert Modeling View Optimize Help

Query Settings

Table.TransformColumnTypes(#"Promoted Headers",{"customerID", type text}, {"gender", type text})

gender SeniorCitizen Partner Dependents tenure

Valid 100% Valid 100% Valid 100% Valid 100% Valid 100% Valid 100% Valid 100%

Error 0% Error 0% Error 0% Error 0% Error 0% Error 0% Error 0%

Empty 0% Empty 0% Empty 0% Empty 0% Empty 0% Empty 0% Empty 0%

1 7590-VHVE Copy

2 5575-GNVD Remove

3 3668-QPVB Remove Other Columns

4 7795-CFOC Duplicate Column

5 9237-HQTL Add Column From Examples...

6 9505-CDSK Remove Duplicates

7 1452-KDQV Remove Errors

8 6713-OKDA Change Type

9 7892-PQOK Transform

10 6388-TABG Replace Values...

11 9763-GRSK Replace Errors...

12 7469-UXBC Split Column

13 8091-TTVA Group By...

14 0280-XGDX Fill

15 5129-JLPS Unpivot Columns

16 3635-SNQY Unpivot Other Columns

17 8131-XWWS Unpivot Only Selected Columns

18 9955-WQFJ Rename...

19 4190-MFLU Move

20 Drill Down

21 Add as New Query

PROPERTIES

Name

Telco_Customer_Churn_Dataset (2)

APPLIED STEPS

Source

Promoted Headers

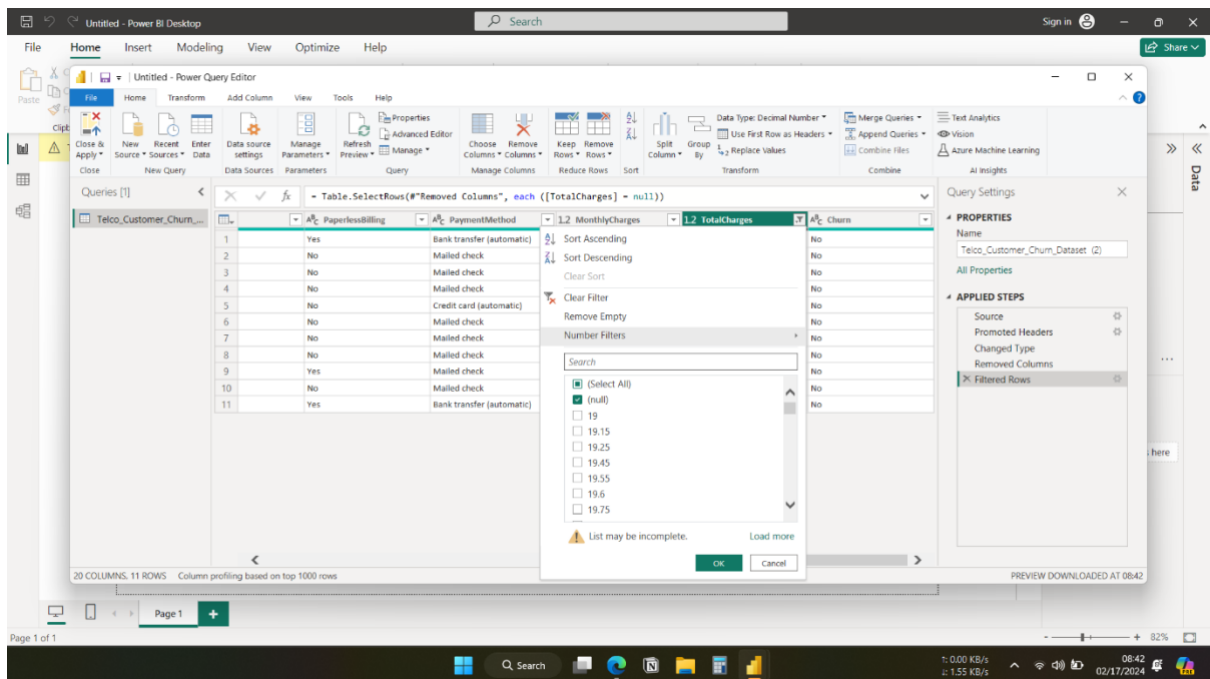
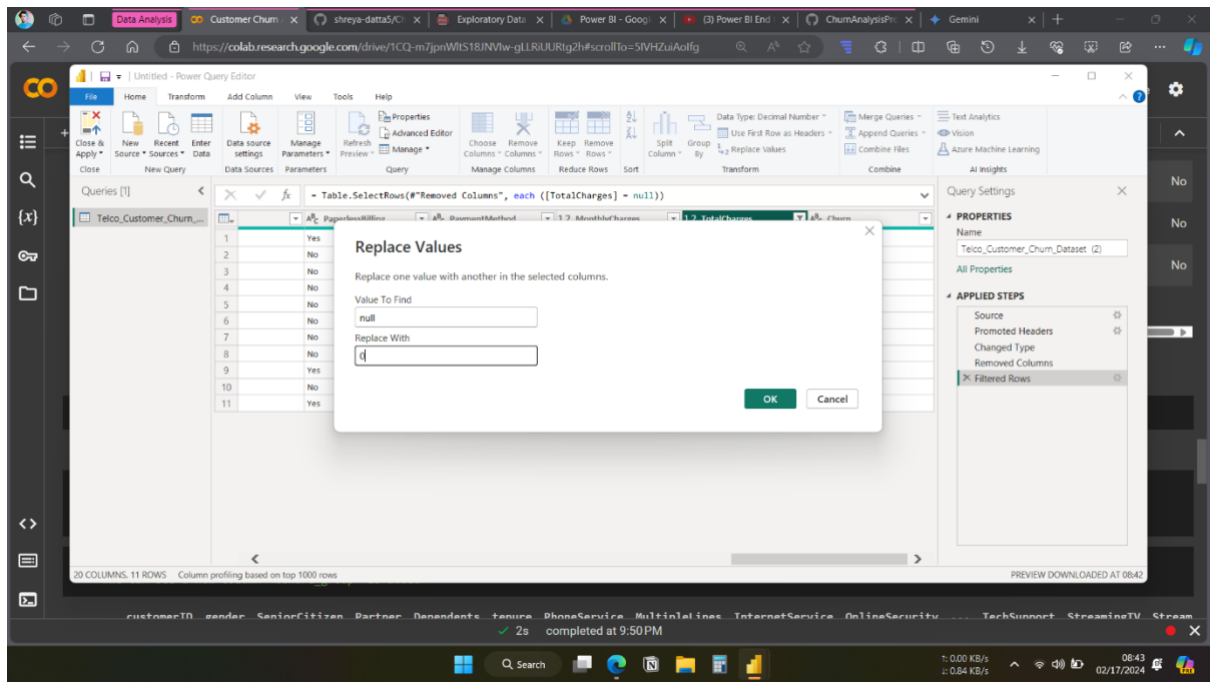
Changed Type

21 COLUMNS, 999+ ROWS Column profiling based on top 1000 rows

PREVIEW DOWNLOADED AT 08:04

Page 1 of 1

82%



Power Query Editor interface showing a data table with columns: PaperlessBilling, PaymentMethod, MonthlyCharges, TotalCharges, and Churn. The table contains 11 rows of data. The interface includes a ribbon with various transformation options (Filter, Add Column, View, Tools, Help) and a right-hand pane for Query Settings (Properties, Applied Steps).

Query Name: Telco_Customer_Churn_Dataset (2)

Applied Steps:

- Source
- Promoted Headers
- Changed Type
- Removed Columns
- Removed Columns
- Replaced Value

Table Data:

	PaperlessBilling	PaymentMethod	MonthlyCharges	TotalCharges	Churn
1	Yes	Bank transfer (automatic)	52.53	0	No
2	No	Mailed check	20.25	0	No
3	No	Mailed check	80.83	0	No
4	No	Mailed check	25.75	0	No
5	No	Credit card (automatic)	56.05	0	No
6	No	Mailed check	19.85	0	No
7	No	Mailed check	25.35	0	No
8	No	Mailed check	20	0	No
9	Yes	Mailed check	19.7	0	No
10	No	Mailed check	73.35	0	No
11	Yes	Bank transfer (automatic)	61.9	0	No

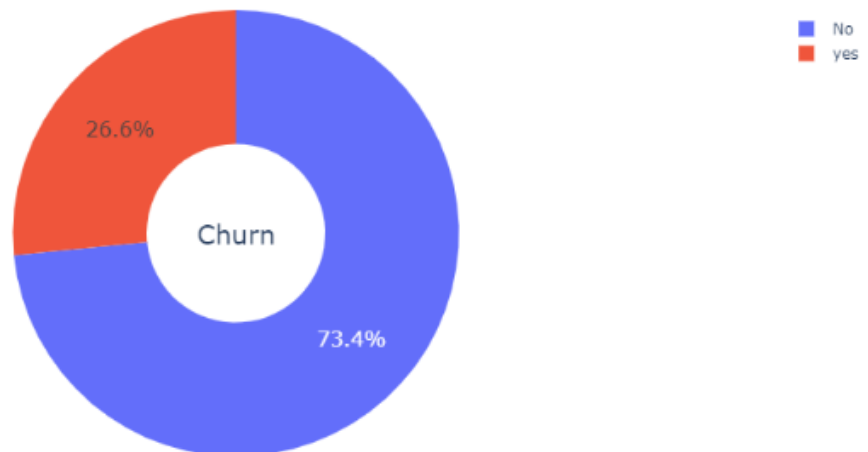
20 COLUMNS, 11 ROWS Column profiling based on top 1000 rows

customerID_gender_SeniorCitizen_Partner_Dependents_tenure_PhoneService_Multiline_InternetService_OnlineSecurity_TechSupport_StreamlineTV_Stream completed at 9:50 PM

Exploratory Data Analysis:

1. Churn distribution

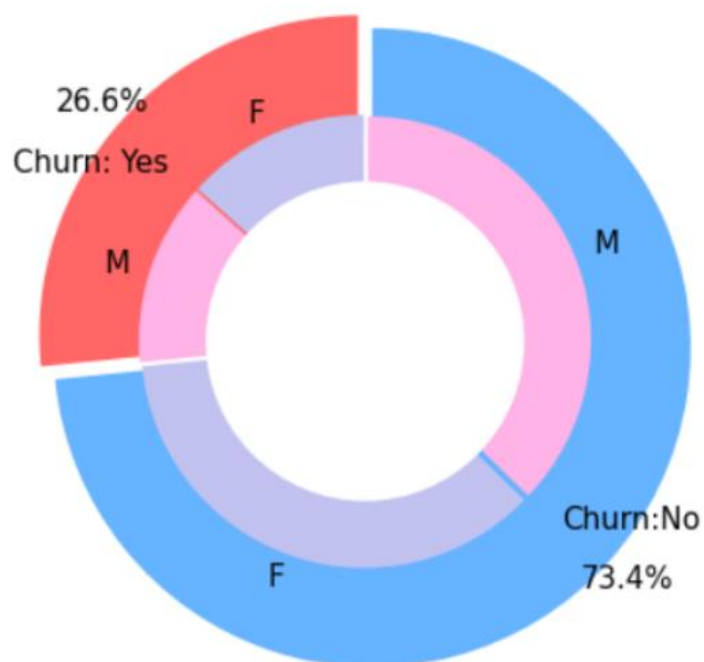
Churn Distributions



26.6 % of customer switched to another firm.

2. churn distribution with respect to gender

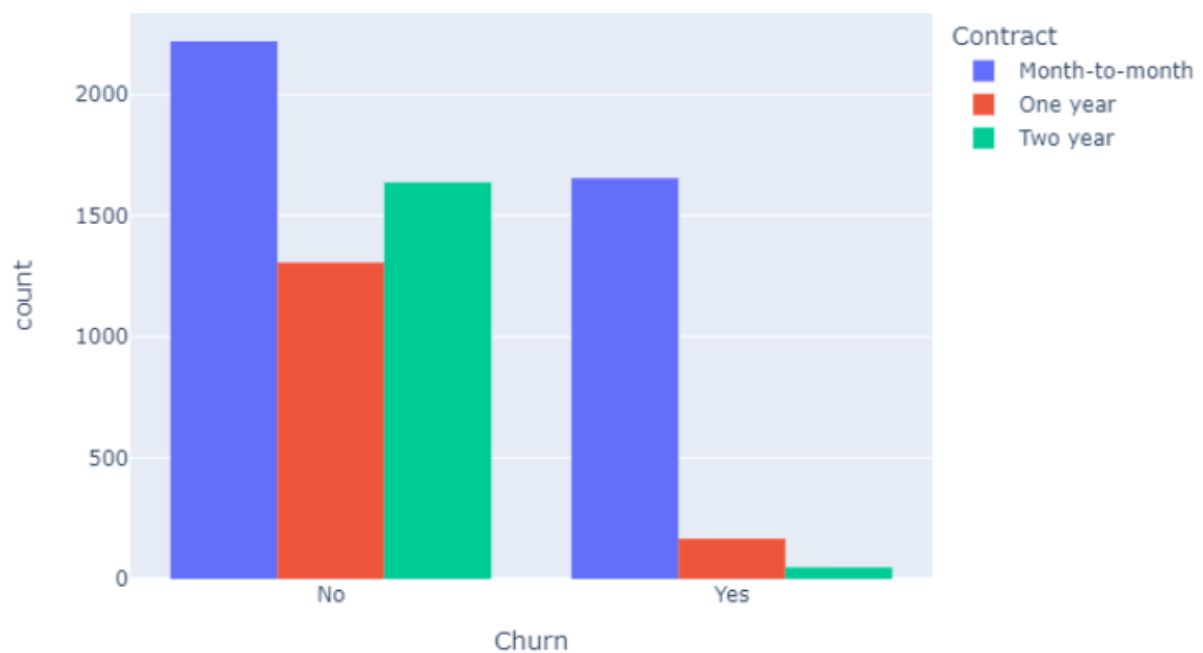
Churn Distribution w.r.t Gender: Male(M), Female(F)



There is negligible difference in customer percentage/count who changed the service provider. Both genders behaved in similar fashion when it comes to migrating to another service provider/firm.

3. Customer contract distribution:

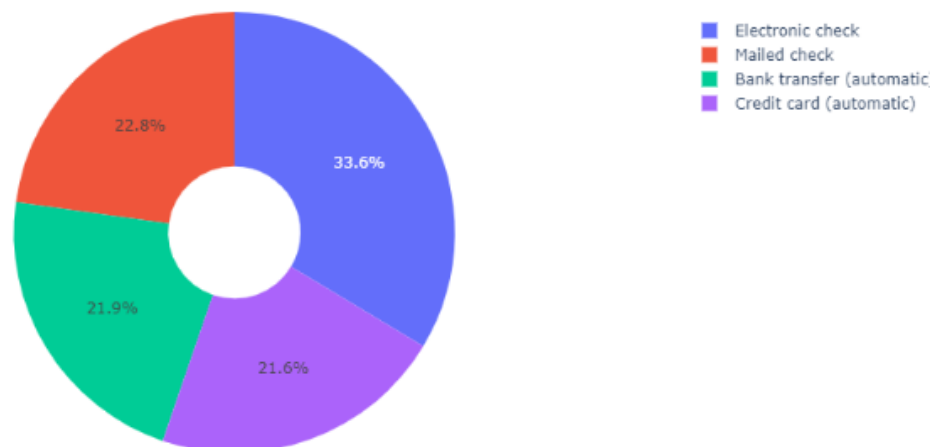
Customer contract distribution



About 75% of customer with Month-to-Month Contract opted to move out as compared to 13% of customers with One Year Contract and 3% with Two Year Contract

4. Payment methods:

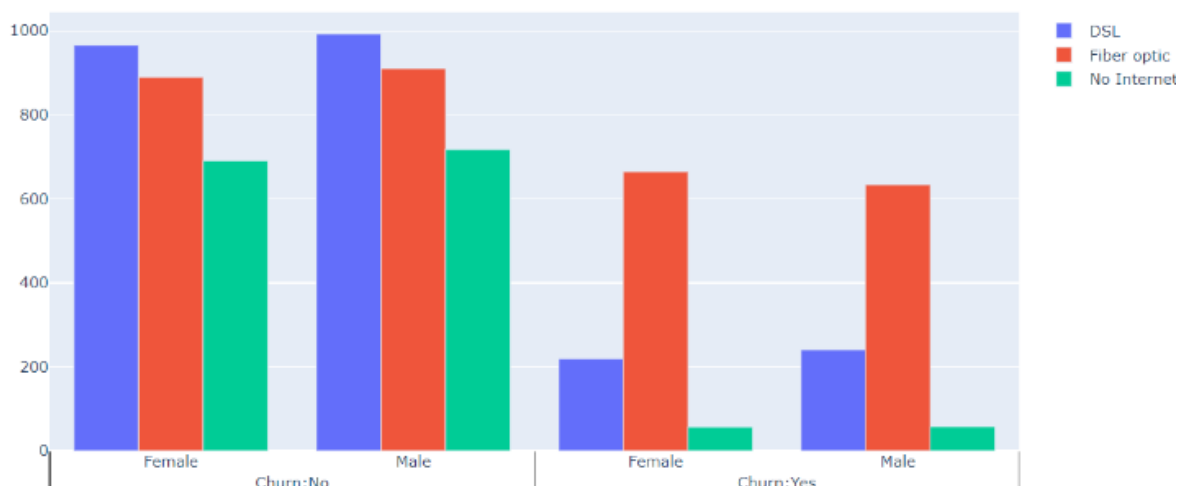
Payment Method Distribution



Major customers who moved out were having Electronic Check as Payment Method. Customers who opted for Credit-Card automatic transfer or Bank Automatic Transfer and Mailed Check as Payment Method were less likely to move out.

5. Internet services:

Churn Distribution w.r.t. Internet Service and Gender

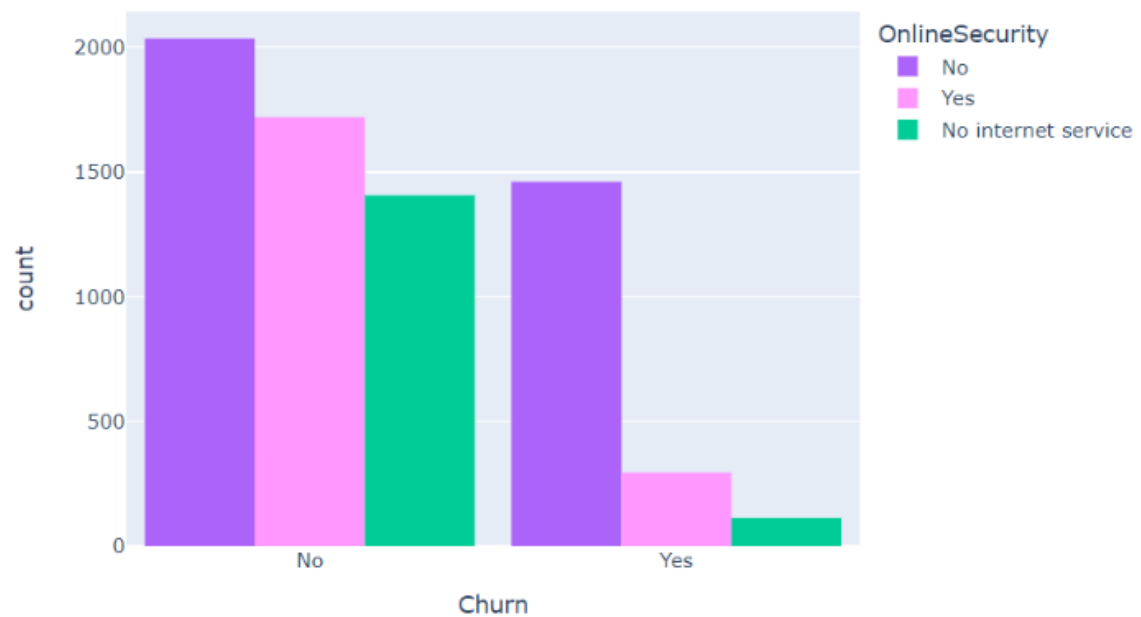


Several customers choose the Fiber optic service and it's also evident that the customers who use Fiber optic have high churn rate, this might suggest a dissatisfaction with this type of internet

service. Customers having DSL service are majority in number and have less churn rate compared to Fibre optic service.

6. Online Security:

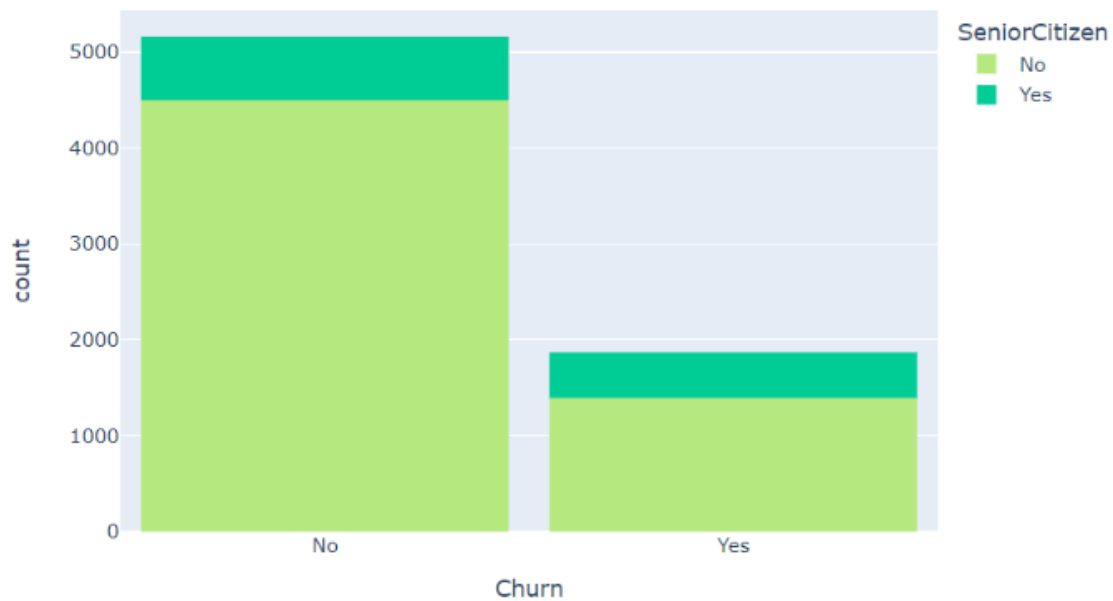
Churn w.r.t Online Security



Majority of churners don't have online security.

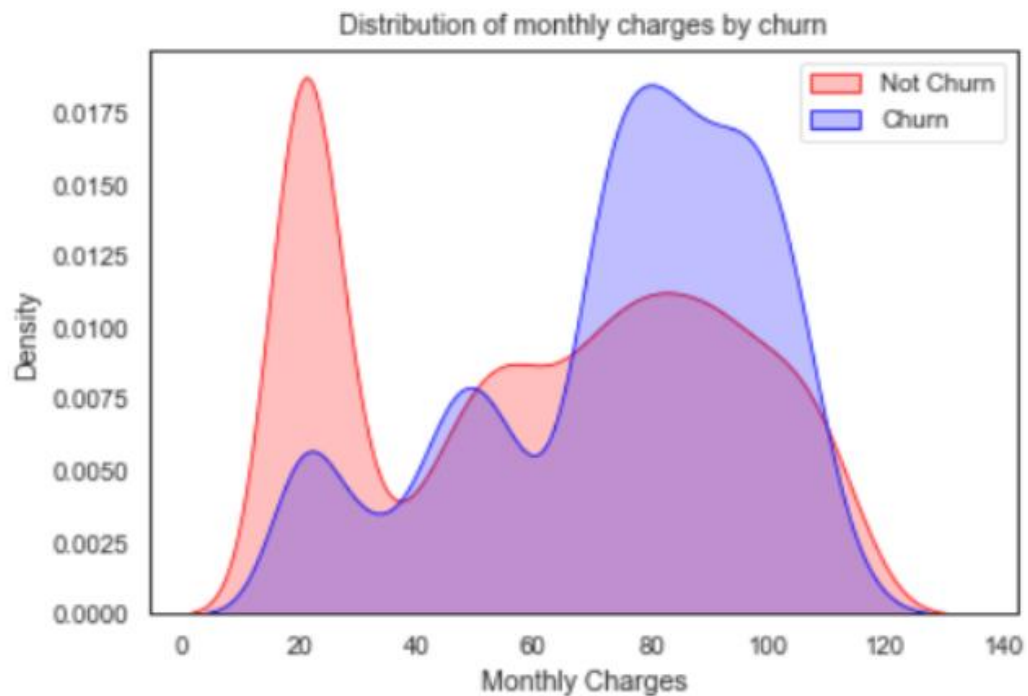
7. Senior Citizen:

Churn distribution w.r.t. Senior Citizen

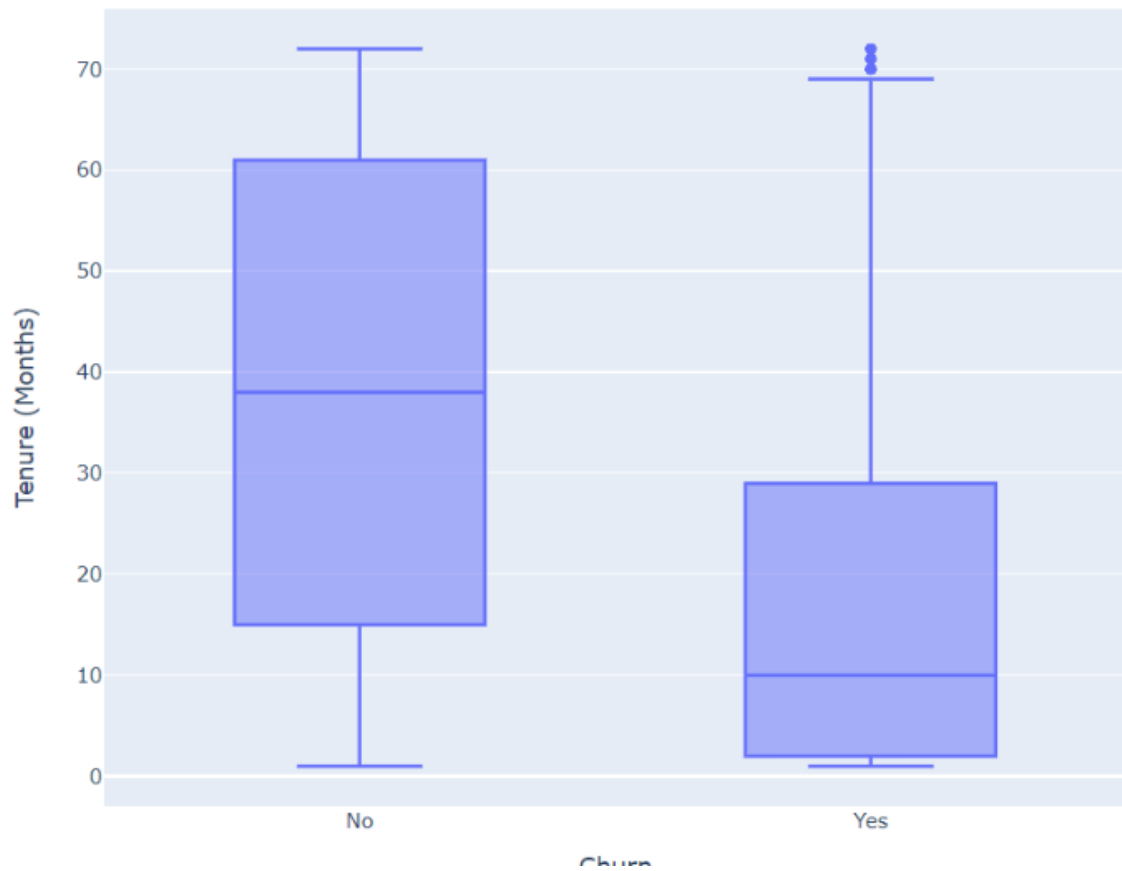


Most of the senior citizens churn; the number of senior citizens are very less in over all customer base.

8. Distribution with Charges and tenure:



Tenure vs Churn



Customers with higher Monthly Charges are also more likely to churn.

New customers are more likely to churn.