# Churn Analysis Report of a Telecom Company (Using SQL)

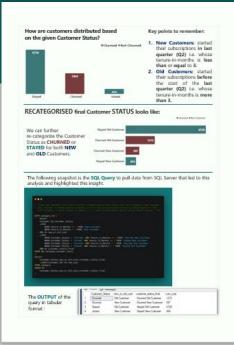
#### **MAIN TOOL USED:**

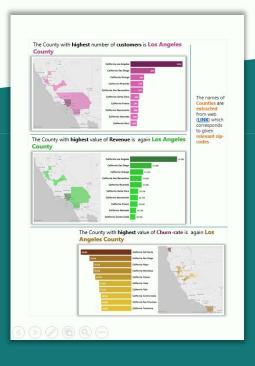
SQL (Analysis)

#### **OTHER TOOLS USED:**

- Excel (web scraping)
- Power BI (visualization)
- Power point (presentation)

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#### PLEASE READ THE FOLLOWING:

- The objective of this SQL project is to present a report (newsletter style) perform some BASIC Exploratory Data Analysis (EDA) of a fictional Telecommunications Company.
- 2. The datasets was taken from Data playground of Maven Analytics (LINK).
- The SQL codes used in this project are to be run in SQL-server.
- 4. The server is hosted on a **localhost** and a **database** named **"TelecomCustomerChurn"** is created using SSMS GUI.
- 5. The **two csv files** are imported into this database as tables and created **two corresponding tables** named:
  - "telecom\_customer\_churn" and "telecom\_zipcode\_population."
- 6. Also, the names of **Counties** are **extracted** from web (**LINK**) which corresponds to given **relevant zip-codes** in one of the data sets as follows:

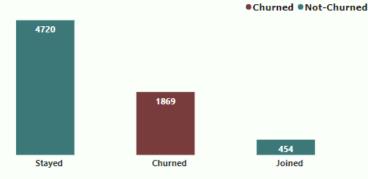
#### Get data > From Other Sources > From Web

7. For **better understanding** of the outputs, they are **visualized** in **Power BI** by extracting the outputs as follows:

# Get data > SQL Server > import (Advanced Options:SQL Statements) > OK

- 8. Finally the codes and their respective visuals are organized in **powerpoint** and then, saved as **pdf file**.
- 9. Feedbacks and suggestions are always appreciated. THANK YOU.

### How are customers distributed based on the given Customer Status?



#### **Key points to remember:**

- New Customers: started their subscriptions in last quarter (Q2) i.e. whose tenure-in-months is less than or equal to 3.
- 2. Old Customers: started their subscriptions before the start of the last quarter (Q2) i.e. whose tenure-in-months is more than 3.

#### **RECATEGORISED** final Customer **STATUS** looks like:

We can further re-categorize the Customer Status as **CHURNED** or **STAYED** for both **NEW** and **OLD** Customers.



The following snapshot is the **SQL Query** to pull data from SQL Server that led to this analysis and highlighted this insight.

```
-- Count the customer with status further categorised by both status and 'Old Customer'/'New Customer'
-- New Customer -> whose tenure in months is <= 3 (who started subscription in latest quarter - Q2)
-- Old Customer -> whose tenure in months is > 3 (who started subscription before latest quarter - Q2)

WITH category AS (
    SELECT
    Customer_ID,Customer_Status
    ,CASE
        WHEN Tenure_in_Months <= 3 THEN 'New Customer'
        WHEN Tenure_in_Months > 3 THEN 'Old Customer'
        END AS new_or_old_cust
    ,CASE
        WHEN Customer_Status = 'Churned' AND Tenure_in_Months <= 3 THEN 'Churned New Customer'
        WHEN Customer_Status = 'Joined' AND Tenure_in_Months <= 3 THEN 'Stayed New Customer'
        WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 'Churned Old Customer'
        WHEN Customer_Status = 'Stayed' AND Tenure_in_Months > 3 THEN 'Stayed Old Customer'
        WHEN Customer_status_final
        FROM dbo.telecom_customer_churn
)

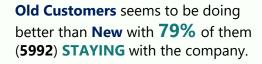
SELECT
    Customer_Status,new_or_old_cust,customer_status_final
        ,COUNT(Customer_ID) AS num_cust

FROM category
    GROUP BY
    Customer_Status,new_or_old_cust,customer_status_final
    ;

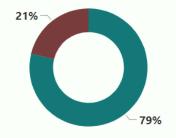
i
```

	Treating	ges		
	Customer_Status	new_or_old_cust	customer_status_final	num_cust
1	Chumed	Old Customer	Chumed Old Customer	1272
2	Chumed	New Customer	Chumed New Customer	597
3	Stayed	Old Customer	Stayed Old Customer	4720
4	Joined	New Customer	Stayed New Customer	454

**1051 New Customers** started subscribing in **Q2** and **597 (57%)** of them **CHURNED** 







Stayed New Customer ● Churned New Customer

● Stayed Old Customer ● Churned Old Customer

The **Churned New Customers** seemed to have inflated the **OVERALL CHURN-RATE** a bit in **Q2**.

26.54%



Churned Not-Churned

```
The percentage distribution of stayed and churned NEW customers

SELECT

ROUND(

108.9 * COUNT(CASE WHEN Customer_Status = 'Joined' AND Tenure_in_Months <= 3 THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Tenure_in_Months <= 3 THEN 1 ELSE NULL END)

/B AS stayed_NEW_to_all_new_percent

/ROUND(

108.9 * COUNT(CASE WHEN Customer_Status = 'Churned' AND Tenure_in_Months <= 3 THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Tenure_in_Months <= 3 THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Tenure_in_Months <= 3 THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Customer_Status = 'Stayed' AND Tenure_in_Months > 3 THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Tenure_in_Months > 3 THEN 1 ELSE NULL END)

/B AS stayed_DLD_to_all_new_percent

/ROUND(

109.0 * COUNT(CASE WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Tenure_in_Months > 3 THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Tenure_in_Months > 3 THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Customer_Status = 'Churned' THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Customer_Status = 'Churned' THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Customer_Status = 'Churned' THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Customer_Status = 'Churned' THEN 1 ELSE NULL END)

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/COUNT(CASE WHEN Customer_Status = 'Churned' THEN 1 ELSE NULL END)

/COUNT(CASE WHEN Customer_Status = 'Churned' THEN 1 ELSE NULL END)
```

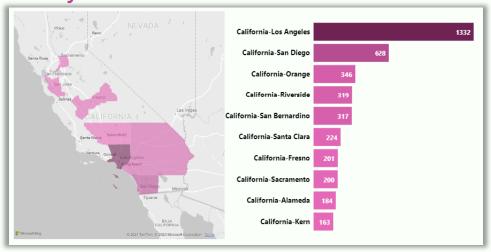
The snapshot to the left is the SQL Queries to pull data from SQL Server that led to this analysis and highlighted this insight.

The **OUTPUTS** of the queries in tabular formats:

chum\_rate\_Q2 1 26.540000000000

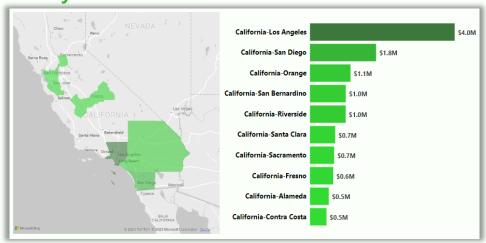
	stayed_NEW_to_all_new_percent	chumed_NEW_to_all_new_percent
1	43.00000000000	57.00000000000
	stayed_OLD_to_all_new_percent	chumed_OLD_to_all_new_percent
1	79.00000000000	21.000000000000

# The County with **highest** number of **customers** is **Los Angeles County**



The names of Counties are extracted from web (LINK) which corresponds to given relevant zipcodes

# The County with **highest** value of **Revenue** is again **Los Angeles County**



#### The County with highest value of Churn-rate is Del Norte



The snapshot given just below this texts is the **SQL Queries** to pull data from SQL Server that led to this analysis and insight on the reasons for each churn category.

```
TOP 10 (zip.state_county), COUNT(cust.Customer_ID) AS num_customer
   dbo.telecom_customer_churn AS cust
   LEFT JOIN dbo.telecom_zipcode_population AS zip
   ON cust.Zip_Code = zip.Zip_Code
GROUP BY zip.state_county
ORDER BY COUNT(cust.Customer_ID) DESC
SELECT
    TOP 10 (zip.state_county), ROUND(SUM(cust.Total_Revenue), 2) as total_revenue
FROM
   dbo.telecom_customer_churn AS cust
   LEFT JOIN dbo.telecom_zipcode_population AS zip
   ON cust.Zip_Code = zip.Zip_Code
GROUP BY zip.state_county
ORDER BY SUM(cust.Total_Revenue) DESC
SELECT
    TOP 10 zip.state_county
        100.0*COUNT(CASE WHEN cust.Customer_Status = 'Churned' THEN 1 ELSE NULL END)
        /COUNT(cust.Customer_ID)
    ,1) AS churn_rate
FROM
    dbo.telecom_customer_churn AS cust
    LEFT JOIN dbo.telecom_zipcode_population AS zip
   ON cust.Zip_Code = zip.Zip_Code
GROUP BY zip.state_county
order by ROUND(
        100.0*COUNT(CASE WHEN cust.Customer_Status = 'Churned' THEN 1 ELSE NULL END)
       /COUNT(cust.Customer_ID)
    ,1) DESC
```

The **OUTPUT** of the queries are just below in tabular format as follows:

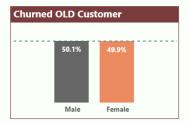
state_county	num_customer
California-Los Angeles	1332
California-San Diego	628
California-Orange	346
California-Riverside	319
California-San Bernardino	317
California-Santa Clara	224
California-Fresno	201
California-Sacramento	200
California-Alameda	184
California-Kem	163

state_county	total_revenue
California-Los Angeles	3991167.56
California-San Diego	1823176.17
California-Orange	1118284.55
California-San Bernardino	986738.53
California-Riverside	984698.7
California-Santa Clara	695409.88
California-Sacramento	665374.7
California-Fresno	641268.54
California-Alameda	517882.81
California-Contra Costa	455454.66

state_county	chum_rate
Califomia-Del Norte	50.000000000000
California-San Diego	41.4000000000000
California-Stanislaus	37.5000000000000
California-Napa	37.500000000000
California-Colusa	35.7000000000000
Califomia-Yuba	31.800000000000
California-Yolo	31.7000000000000
California-Contra Costa	30.100000000000
California-San Francisco	29.800000000000
California-Tuolumne	29.500000000000

**Gender proportions** are almost **BALANCED** in all the groups of customers based on recategorized customer status.



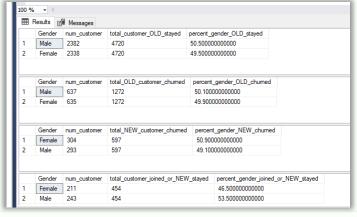




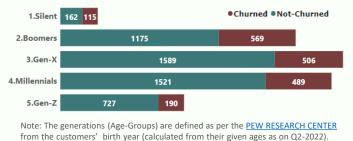




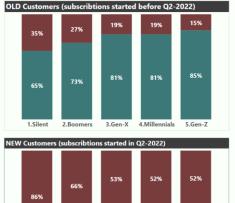
The snapshot to the left is the SQL Queries to pull data from SQL Server that led to this analysis and highlighted this insight.



### Overall Distribution of Customers based on **Generations** (Age-Group):



- Majority of the customers belong to boomers, generation-X and millennials.
- As observed from the both stacked charts to the right, the customers of OLDER GENERATIONS (Age-Groups) have higher churn-rates.



3.Gen-X

14%

2.Boomers

48%

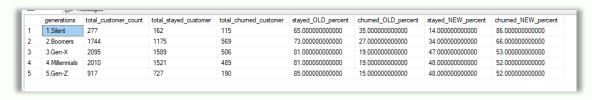
4.Millennials

48%

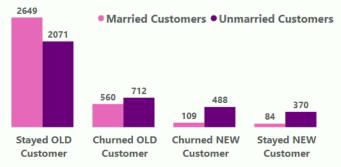
5.Gen-Z

The snapshot below is the **SQL Queries** to pull data from SQL Server that led to this analysis and highlighted this insight.

```
Churn-distribution based on Age-Group
WITH ctetable AS {
    SELECT
    Customer_ID_Customer_Status = 'Stayed' AND Tenure_in_Months > 3 THEN 'Stayed OLD Customer'
    WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 'Stayed OLD Customer'
    WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 'Churned OLD Customer'
    WHEN Customer_Status = 'Churned' AND Tenure_in_Months >= 3 THEN 'Churned NEW Customer'
    WHEN Customer_Status = 'Churned' AND Tenure_in_Months >= 3 THEN 'Churned NEW Customer'
    WHEN Age BETWEEN 10 AND 105 THEN 'S.Gen-Z'
    WHEN Age BETWEEN 10 AND 10 THEN '13.Gen-X'
    WHEN Age BETWEEN 10 AND 15 THEN '13.Gen-X'
    WHEN Age BETWEEN 10 AND 11 THEN '14.Millennials'
    WHEN Age BETWEEN 10 AND 11 THEN '13.Gen-X'
    WHEN Age BETWEEN 10 AND 11 THEN '13.Gen-X'
    WHEN Age BETWEEN 10 AND 10 THEN '13.Gen-X'
    WHEN Age BETWEEN 10 THE
```

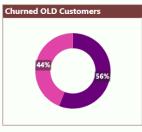


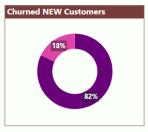
# Distribution of **Married** and **Unmarried** customers within each **Recategorized Customer Status:**



- ONLY in Stayed-OLD-Customers group, the married customers are more than unmarried.
- Majority of the NEW customers (both churned and stayed) who started their subscriptions in the latest quarter (Q2-2022) are unmarried









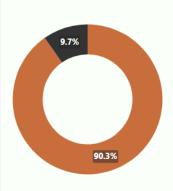
The snapshot below is the **SQL Queries** to pull data from SQL Server that led to this analysis and highlighted this insight.

```
. .
WITH ctetable AS (
     SELECT
           Customer_ID, Gender, Married, Number_of_Dependents, Tenure_in_Months, Customer_Status
           .CASE
                 WHEN Tenure_in_Months > 3 AND Customer_Status = 'Stayed' THEN 'Stayed OLD Customer'
                 WHEN Tenure_in_Months > 3 AND Customer_Status = 'Churned' THEN 'Churned OLD Customer'
WHEN Tenure_in_Months <= 3 AND Customer_Status = 'Joined' THEN 'Stayed NEW Customer'
WHEN Tenure_in_Months <= 3 AND Customer_Status = 'Churned' THEN 'Churned NEW Customer'
           END AS final_customer_status
     FROM dbo.telecom_customer_churn
SELECT Customer_Status, final_customer_status
      ,SUM(CASE WHEN Married = 1 THEN 1 ELSE NULL END) AS married
,SUM(CASE WHEN Married = 0 THEN 1 ELSE NULL END) AS unmarried
,ROUND(100.0 * SUM(CASE WHEN Married = 1 THEN 1 ELSE NULL END)/COUNT(Customer_ID)
            ,0) AS married_percent
      ,ROUND(100.0 * SUM(CASE WHEN Married = 0 THEN 1 ELSE NULL END)/COUNT(Customer_ID)
           ,0) AS unmarried_percent
FROM ctetable
GROUP BY Customer_Status, final_customer_status
```

	Customer_Status	final_customer_status	married	unmarried	mamed_percent	unmarried_percent
1	Chumed	Chumed NEW Customer	109	488	18.000000000000	82.000000000000
2	Chumed	Chumed OLD Customer	560	712	44.0000000000000	56.000000000000
3	Joined	Stayed NEW Customer	84	370	19.000000000000	81.000000000000
4	Stayed	Stayed OLD Customer	2649	2071	56.000000000000	44.0000000000000

# 9 in 10 customers (90%) subscribed to phone services.

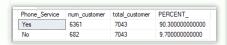




Among them, the percentage of **multiple lines subscribed CHURNED CUSTOMERS** were than the **STAYED CUSTOMERS** counterparts.



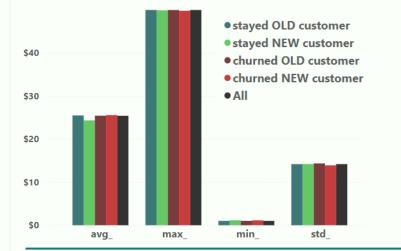
The snapshot below is the **SQL Queries** to pull data from SQL Server that led to this analysis and highlighted this insight.



	final_customer_status	Multiple_Lines	num_customer_phone_service_yes	percent_multiple_lines_within_status
1	chumed NEW customer	Yes	128	24.200000000000
2	chumed NEW customer	No	401	75.800000000000
3	chumed OLD customer	No	448	38.30000000000
4	chumed OLD customer	Yes	722	61.700000000000
5	stayed NEW customer	Yes	45	10.80000000000
6	stayed NEW customer	No	371	89.200000000000
7	stayed OLD customer	No	2170	51.100000000000
8	stayed OLD customer	Yes	2076	48.900000000000

The distributions of **VARIOUS STATISTICS** of **monthly long-distance charges ACROSS** the **CUSTOMER STATUS** are almost same with that of the values of overall **POPULATION**.

Overall (All) Monthly Charges:



The snapshot below is the **SQL Queries** to pull data from SQL Server that led to this analysis and highlighted this insight.

```
- analysis of Avg_Monthly_Long_Distance_Charges based on final status

WITH ctetable AS

(SELECT Customer_ID,Customer_Status
,CASE

WHEN Customer_Status = 'Stayed' AND Tenure_in_Months > 3 THEN 'stayed OLD customer'
WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 'churned OLD customer'
WHEN Customer_Status = 'Joined' AND Tenure_in_Months <= 3 THEN 'churned NEW customer'
WHEN Customer_Status = 'Joined' AND Tenure_in_Months <= 3 THEN 'stayed NEW customer'
ELSE NULL

END AS final_customer_status, Phone_Service, Avg_Monthly_Long_Distance_Charges, Multiple_Lines
FROM dbo. telecom_customer_status, ROUND(AVG(Avg_Monthly_Long_Distance_Charges), 2) avg_
ROUND(SMT(VAR(Avg_Monthly_Long_Distance_Charges), 2) min_
ROUND(MAX(Avg_Monthly_Long_Distance_Charges), 2) min_
ROUND(MAX(Avg_Monthly_Long_Distance_Charges), 2) min_
FROM tetable
GROUP BY final_customer_status

UNION ALL

-- overall Avg_Monthly_Long_Distance_Charges

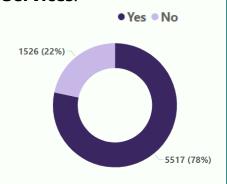
**ELECT**

All' AS final_customer_status, ROUND(AVG(Avg_Monthly_Long_Distance_Charges), 2) avg_
ROUND(SORT(VAR(Avg_Monthly_Long_Distance_Charges)), 2) std_,
ROUND(SORT(VAR(Avg_Monthly_Long_Distance_Charges)), 2) min_
FROM dbo.telecom_customer_churn

;
```

	final_customer_status	avg_	std_	max_	min_
1	stayed OLD customer	25.5	14.19	49.99	1.01
2	chumed OLD customer	25.44	14.36	49.98	1.02
3	stayed NEW customer	24.34	14.2	49.91	1.15
4	chumed NEW customer	25.58	13.91	49.81	1.13
5	All	25.42	14.2	49.99	1.01

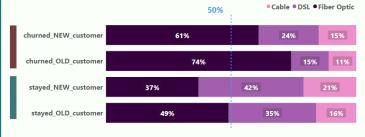
Almost 8 in 10 customers (78%) subscribed to Internet Services.



Majority of the customers who have **CHURNED subscribed** to internet services through the type **FIBRE-OPTIC**.

FIBRE

- 61% of churned NEW customers
- 74% of churned OLD customers



The snapshot below is the **SQL Queries** to pull data from SQL Server that led to this analysis and highlighted this insight.

```
-- Internet Services subscriptions

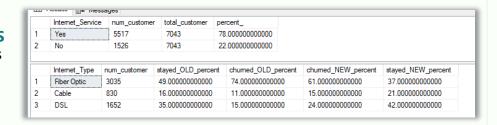
SELECT Internet_Service, COUNT(Customer_ID)/SUM(COUNT(Customer_ID)) over() AS total_customer , ROUND(180.9-COUNT(Customer_ID)/SUM(COUNT(Customer_ID)) over(), 8) AS percent_
FROM dbo.telecom_customer_churn
GROUP PS Internet_Service
;

-- churn-rate analysis based on Internet Type
MTH ctetable AS (
SELECT
Customer_ID,Internet_Service,Internet_Type
, CASE

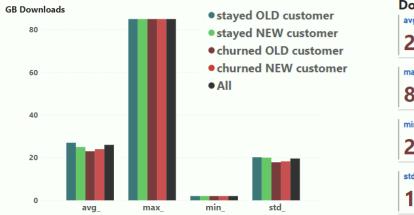
WHEN Customer_Status = 'Stayed' AND Tenure_in_Months > 3 THEN 'stayed_OLD_customer'
WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 'churned_IDD_customer'
WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 'churned_IDD_customer'
WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 'churned_RWZ_customer'
WHEN Customer_Status = 'Churned' AND Tenure_in_Months < 3 THEN 'churned_RWZ_customer'
WHEN Customer_Status = 'Stayed_AND Tenure_in_Months < 3 THEN 'stayed_NEW_customer'

ELSE NOLL END AS final_customer_status
FROM dbo.telecom_customer_churn
}

SELECT
Internet_Type, COUNT(CASE WHEN final_customer
,ROUND_GROUP (CASE WHEN final_customer_status = 'stayed_OLD_customer' THEN I ELSE NOLL END)
/SUM(COUNT(CASE WHEN final_customer_status = 'churned_OLD_customer' THEN I ELSE NOLL END)
/SUM(COUNT(CASE WHEN final_customer_status = 'churned_OLD_customer' THEN I ELSE NOLL END)
/SUM(COUNT(CASE WHEN final_customer_status = 'churned_NW_customer' THEN I ELSE NOLL END)
/SUM(COUNT(CASE WHEN final_customer_status = 'churned_NW_customer' THEN I ELSE NOLL END)
/SUM(COUNT(CASE WHEN final_customer_status = 'churned_NW_customer' THEN I ELSE NOLL END)
/SUM(COUNT(CASE WHEN final_customer_status = 'churned_NW_customer' THEN I ELSE NOLL END)
/SUM(COUNT(CASE WHEN final_customer_status = 'churned_NW_customer' THEN I ELSE NOLL END)
/SUM(COUNT(CASE WHEN final_customer_status = 'churned_NW_customer' THEN I ELSE NOLL END)
/SUM(COUNT(CASE WHEN final_customer_status = 'stayed_NW_customer' THEN I ELSE NOLL END)
/SUM(COUNT(CASE WHEN final_customer_status = 'stayed_NW_customer' THEN I ELSE NOLL END)
/SUM(COUNT(CASE WHEN final_custom
```



# The distributions of **VARIOUS STATISTICS** of **average monthly downloads (GB) ACROSS** the **CUSTOMER STATUS** are almost same with that of the **POPULATION**



Overall (All) Monthly
Downloads (GB):

avg\_
26.00

max\_
85.00

min\_
2.00

std\_
19.59

The snapshot below is the **SQL Queries** to pull data from SQL Server that led to this analysis and highlighted this insight.

```
-- analysis of Avg_Monthly_GB_Download based on final status
WITH ctetable AS (
SELECT

Customer_ID,Avg_Monthly_GB_Download
,CASE

WHEN Customer_Status = 'Stayed' AND Tenure_in_Months > 3 THEN 'stayed_OLD_customer'
WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 'churned_OLD_customer'
WHEN Customer_Status = 'Churned' AND Tenure_in_Months <= 3 THEN 'churned_OLD_customer'
WHEN Customer_Status = 'Joined' AND Tenure_in_Months <= 3 THEN 'churned_NEW_customer'
ELSE NULL END AS final_customer_status
FROM dbo.telecom_customer_churn
WHERE Internet_Service = 'Yes'
)
SELECT

final_customer_status, MAX(Avg_Monthly_GB_Download) max_,MIN(Avg_Monthly_GB_Download) min_,AVG(Avg_Monthly_GB_Download) avg_,ROUND(SQRT(VAR(Avg_Monthly_GB_Download)),2) AS std_
FROM ctetable
GROUP BY final_customer_status

UNION ALL

-- overall Avg_Monthly_GB_Download

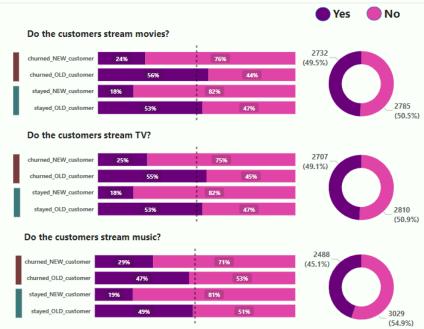
SELECT

'All (Overall)' AS final_customer_status,MAX(Avg_Monthly_GB_Download)
max_,MIN(Avg_Monthly_GB_Download) min_
,AVG(Avg_Monthly_GB_Download) avg_, ROUND(SQRT(VAR(Avg_Monthly_GB_Download)),2) AS std_
FROM dbo.telecom_customer_churn
;
```

	final_customer_status	max_	min_	avg_	std_
1	stayed_NEW_customer	85	2	25	20.01
2	chumed_OLD_customer	85	2	23	17.84
3	stayed_OLD_customer	85	2	27	20.19
4	chumed_NEW_customer	85	2	24	18.27
5	All (Overall)	85	2	26	19.59

Among the INTERNET SUBSCRIBERS, the distributions of the streaming-customers across the CHURNED and STAYED statuses are similar, and

The overall streaming status **does not** seem to influence the **churn behaviour** 



```
Count (CASE WHEN Streaming Movies = 'Yes' THEN 1 ELSE NULL END) AS movies_yes

COUNT(CASE WHEN Streaming_Movies = 'No' THEN 1 ELSE NULL END) AS movies_no

COUNT(CASE WHEN Streaming_Movies = 'Yes' THEN 1 ELSE NULL END) AS movies_no

COUNT(CASE WHEN Streaming_Music = 'Yes' THEN 1 ELSE NULL END) AS music_no

COUNT(CASE WHEN Streaming_TW = 'Yes' THEN 1 ELSE NULL END) AS music_no

COUNT(CASE WHEN Streaming_TV = 'Yes' THEN 1 ELSE NULL END) AS tv_yes

COUNT(CASE WHEN Streaming_TV = 'Yes' THEN 1 ELSE NULL END) AS tv_yes

COUNT(CASE WHEN Streaming_TV = 'Yes' THEN 1 ELSE NULL END) AS tv_no

COUNT(CASE WHEN Streaming_TV = 'Yes' THEN 1 ELSE NULL END) AS tv_no

CUStomer_TO, Internet_Service = 'Yes'

**Churn_rate analysis based on Streaming_Movies, Streaming_Music, Streaming_TV

CASE

WHEN Customer_Status = 'Stayed' AND Tenure_in_Months > 3 THEN 'stayed_OLD_customer'

WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 'churned_OLD_customer'

WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 'churned_NEW_customer'

WHEN Customer_Status = 'Churned' AND Tenure_in_Months < 3 THEN 'stayed_NEW_customer'

WHEN Customer_Status = 'Oloned' AND Tenure_in_Months < 3 THEN 'stayed_NEW_customer'

ELSE WULL END AS final_customer_status

FROM dbo.telecom_customer_churn

**)

**SECET**

final_customer_status

ROUNG(100_0*COUNT(CASE WHEN Streaming_Movies = 'Yes' THEN 1 ELSE NULL END)/COUNT(*)

**,0 AS movies_no

ROUNG(100_0*COUNT(CASE WHEN Streaming_Movies = 'No' THEN 1 ELSE NULL END)/COUNT(*)

**,0 AS movies_no

ROUNG(100_0*COUNT(CASE WHEN Streaming_TV = 'Yes' THEN 1 ELSE NULL END)/COUNT(*)

**,0 As movies_no

ROUNG(100_0*COUNT(CASE WHEN Streaming_TV = 'Yes' THEN 1 ELSE NULL END)/COUNT(*)

**,0 As movies_no

ROUNG(100_0*COUNT(CASE WHEN Streaming_TV = 'Yes' THEN 1 ELSE NULL END)/COUNT(*)

**,0 As movies_no

ROUNG(100_0*COUNT(CASE WHEN Streaming_TV = 'Yes' THEN 1 ELSE NULL END)/COUNT(*)

**,0 As movies_no

ROUNG(100_0*COUNT(CASE WHEN Streaming_TV = 'Yes' THEN 1 ELSE NULL END)/COUNT(*)

**,0 As movies_no

ROUNG(100_0*COU
```

The snapshot to the left is the **SQL Queries** to pull data from SQL Server that led to this analysis and highlighted this insight.

novies_yes movies_no	music_yes music	_no tv_yes	tv_no	total			
2732 2785	2488 3029	2707	2810	5517			
final_customer_status	movies_yes	movies_no		music_yes	music_no	TV_yes	TV_no
stayed_NEW_customer	18.000000000000	82.00000000	0000	19.000000000000	81.000000000000	18.000000000000	82.00000000000
chumed_OLD_customer	56.000000000000	44.00000000	0000	47.0000000000000	53.000000000000	55.000000000000	45.000000000000
stayed_OLD_customer	53.000000000000	47.00000000	0000	49.000000000000	51.000000000000	53.000000000000	47.000000000000
chumed NEW customer	24.0000000000000	76.00000000	0000	29.000000000000	71.000000000000	25.000000000000	75.00000000000



The snapshot below is the **SQL Queries** to pull data from SQL Server that led to this analysis and highlighted this insight.

```
. . .
WITH ctetable AS (
             \label{local_continuity} Customer_ID, Online\_Security, Online\_Backup, Device\_Protection\_Plan, Premium\_Tech\_Support, Streaming\_Movies, Streaming\_Music, Streaming\_TV, Unlimited\_Data
              .CASE
                   WHEN Customer_Status = 'Stayed' AND Tenure_in_Months > 3 THEN 'stayed_OLD_customer'
WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 'churned_OLD_customer'
WHEN Customer_Status = 'Churned' AND Tenure_in_Months <= 3 THEN 'churned_NEW_customer'
WHEN Customer_Status = 'Joined' AND Tenure_in_Months <= 3 THEN 'stayed_NEW_customer'
      ELSE NULL END AS final_customer_status
FROM dbo.telecom_customer_churn
       WHERE Internet_Service =
SELECT final_customer_status
ROUND(100.0*COUNT(CASE WHEN Online_Security = 'No' THEN 1 ELSE NULL END)/COUNT(*),0) security_no
ROUND(100.0*COUNT(CASE WHEN Online_Backup = 'Yes' THEN 1 ELSE NULL END)/COUNT(*),0) backup_yes
ROUND(100.0*COUNT(CASE WHEN Online_Backup = 'No' THEN 1 ELSE NULL END)/COUNT(*),0) backup_no
ROUND(100.0*COUNT(CASE WHEN Device_Protection_Plan = 'Yes' THEN 1 ELSE NULL END)/COUNT(*),0)
       , ROUND(100.0*COUNT(CASE WHEN Device_Protection_Plan = 'No' THEN 1 ELSE NULL END)/COUNT(*),0)
protection no
       , ROUND(100.0*COUNT(CASE WHEN Premium_Tech_Support = 'Yes' THEN 1 ELSE NULL END)/COUNT(*),0)
      , ROUND(100.0*COUNT(CASE WHEN Premium_Tech_Support = 'No' THEN 1 ELSE NULL END)/COUNT(*),0)
support_no
FROM ctetable GROUP BY final_customer_status
ORDER BY final_customer_status DESC
```

Ele messages								
final_customer_status	security_yes	security_no	backup_yes	backup_no	protection_yes	protection_no	support_yes	support_no
stayed_OLD_customer	48.0000000000000	52.000000000000	53.000000000000	47.0000000000000	53.000000000000	47.000000000000	48.000000000000	52.000000000000
stayed_NEW_customer	17.0000000000000	83.000000000000	20.000000000000	80.000000000000	12.000000000000	88.000000000000	17.0000000000000	83.000000000000
chumed_OLD_customer	21.0000000000000	79.000000000000	37.0000000000000	63.0000000000000	38.000000000000	62.000000000000	22.000000000000	78.0000000000000
chumed_NEW_customer	7.000000000000	93.000000000000	13.000000000000	87.0000000000000	14.0000000000000	86.000000000000	7.000000000000	93.000000000000

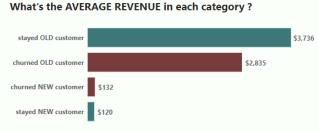
#### \$21.37M Total Revenue

#### As per the **AVERAGE REVENUE**:

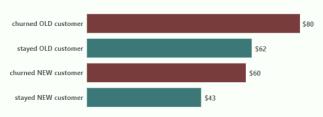
- **1. OLD customers** are **highly valued** regardless of whether they are **CHURNED** or **STAYED**.
- 2. Among the NEW customers the Churned NEW customers have higher average revenue than Stayed NEW customers

As per the AVERAGE MONTHLY

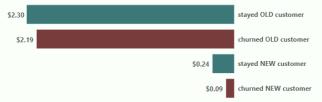
CHARGES: CHURNED customers are
charged higher than STAYED customers
within both OLD and NEW customers







#### What's the AVERAGE REFUND of each category?



#### As per the **AVERAGE REFUND**:

 STAYED customers received higher than CHURNED customers within both OLD and NEW customers.

The following snapshot is the **SQL Query** to pull data from SQL Server that led to this analysis and highlighted this insight.

```
-- analysis of total revenue, monthly charge and refund for each final category
WITH ctetable AS (
SELECT

CASE

WHEN Customer_Status = 'Stayed' AND Tenure_in_Months > 3 THEN 'stayed OLD customer'
WHEN Customer_Status = 'Churned' AND Tenure_in_Months > 3 THEN 'churned OLD customer'
WHEN Customer_Status = 'Churned' AND Tenure_in_Months <= 3 THEN 'churned NEW customer'
WHEN Customer_Status = 'Joined' AND Tenure_in_Months <= 3 THEN 'stayed NEW customer'
ELSE NULL

END AS final_customer_status

,Monthly_Charge,Total_Charges,Total_Revenue,Total_Extra_Data_Charges,Total_Long_Distance_Charges,Total_Refunds
FROM dbo.telecom_customer_churn
)

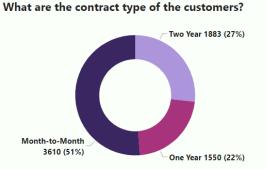
SELECT final_customer_status, ROUND(AVG(Monthly_Charge),2) average_monthly_charge
,ROUND(AVG(Total_Refunds),2) agerage_total_refund
,ROUND(SUM(Total_Revenue),2) AS total_revenue
FROM ctetable
GROUP BY final_customer_status
UNION
SELECT 'All(overall)' AS final_customer_status,ROUND(AVG(Monthly_Charge),2) average_monthly_charge
,ROUND(AVG(Total_Refunds),2) agerage_total_refund
,ROUND(CSUTOtal_Refunds),2) agerage_total_refund
,ROUND(CSUTOtal_Refunds),2) agerage_total_refund
,ROUND(CSUTOtal_Refunds),2) agerage_total_refund
,ROUND(CSUTOtal_Refunds),2) agerage_total_refund
,ROUND(SUM(Total_Revenue),2) AS total_revenue
FROM dbo.telecom_customer_churn
ORDER BY final_customer_status DESC
;
```

<u> </u>				
final_customer_status	average_monthly_charge	average_revenue	agerage_total_refund	total_revenue
stayed OLD customer	61.74	3735.68	2.3	17632392.11
stayed NEW customer	42.78	119.56	0.24	54279.75
chumed OLD customer	79.83	2834.62	2.19	3605638.68
chumed NEW customer	59.54	132.03	0.09	78821.14
All(overall)	63.6	3034.38	1.96	21371131.68

**HALF** of the customers subscribed to services through **Month-to-Month** contract.

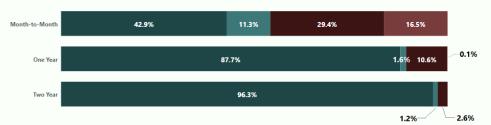
**CHURNED Rate** is higher for customers with **Month-to-Month** Contract

Most of the customers with <u>Two year</u> (Long-term) Contract are stayed OLD customers



Distributions of various final customer statues for every contract type:

● stayed OLD customer ● stayed NEW customer ● churned OLD customer ● churned NEW customer



The following snapshot is the **SQL Query** to pull data from SQL Server that led to this analysis and highlighted this insight.

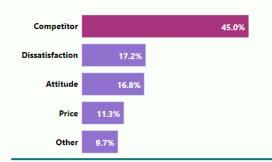
# The **OUTPUT** of the

query in tabular format :

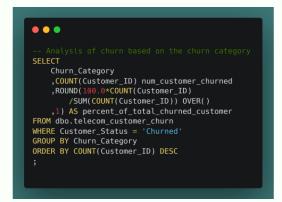
contract_type	num_customer	total_customer	contract_to_overall_percent
Month-to-Month	3610	7043	51.000000000000
One Year	1550	7043	22.000000000000
Two Year	1883	7043	27.000000000000

contract_type	final_customer_status	status_to_contract_type_percent
Month-to-Month	chumed NEW customer	16.480000000000
Month-to-Month	chumed OLD customer	29.36000000000
Month-to-Month	stayed NEW customer	11.30000000000
Month-to-Month	stayed OLD customer	42.850000000000
One Year	chumed NEW customer	0.130000000000
One Year	chumed OLD customer	10.58000000000
One Year	stayed NEW customer	1.550000000000
One Year	stayed OLD customer	87.74000000000
Two Year	chumed OLD customer	2.550000000000
Two Year	stayed NEW customer	1.170000000000
Two Year	stayed OLD customer	96.280000000000

Almost **HALF** of the **churned** customers are because of **COMPETITOR's** influence.



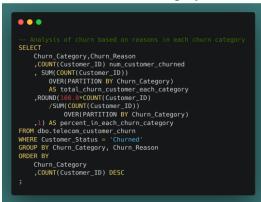
The following snapshot is the **SQL Query** to pull data from SQL Server that led to this analysis and insight on the churn categories.



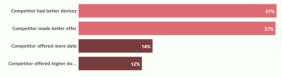
The **OUTPUT** of the above query in tabular format:

Chum_Category	num_customer_chumed	percent_of_total_chumed_customer
Competitor	841	45.000000000000
Dissatisfaction	321	17.200000000000
Attitude	314	16.800000000000
Price	211	11.300000000000
Other	182	9.70000000000

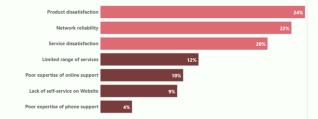
The snapshot at the bottom-left is the **SQL Query** to pull data from SQL Server that led to this analysis and insight on the reasons for each churn category.



## **REASONS** for **churn** based on **COMPETITOR's** influence:



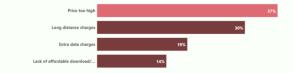
### **REASONS** for **churn** based on **DISSATISFACTION**:



**REASONS** for **churn** based on **ATTITUDE** of the employees:



**REASONS** for **churn** based on **PRICE** of the services:



**Customers churned** for REASONS **OTHER** than the specified:



The **OUTPUT** of the query to the left is in tabular format as follows:

Chum_Category	Chum_Reason	num_customer_chumed	total_chum_customer_each_category	percent_in_each_chum_category
Attitude	Attitude of support person	220	314	70.100000000000
Attitude	Attitude of service provider	94	314	29.90000000000
Competitor	Competitor had better devices	313	841	37.200000000000
Competitor	Competitor made better offer	311	841	37.000000000000
Competitor	Competitor offered more data	117	841	13.900000000000
Competitor	Competitor offered higher download speeds	100	841	11.900000000000
Dissatisfaction	Product dissatisfaction	77	321	24.000000000000
Dissatisfaction	Network reliability	72	321	22.400000000000
Dissatisfaction	Service dissatisfaction	63	321	19.600000000000
Dissatisfaction	Limited range of services	37	321	11.500000000000
Dissatisfaction	Poor expertise of online support	31	321	9.700000000000
Dissatisfaction	Lack of self-service on Website	29	321	9.000000000000
Dissatisfaction	Poor expertise of phone support	12	321	3.700000000000
Other	Don't know	130	182	71.400000000000
Other	Moved	46	182	25.300000000000
Other	Deceased	6	182	3.30000000000
Price	Price too high	78	211	37.000000000000
Price	Long distance charges	64	211	30.300000000000
Price	Extra data charges	39	211	18.500000000000
Price	Lack of affordable download/upload speed	30	211	14.200000000000

# Thank You

By: Yaipharen PM Potsangbam