

Video Streaming App Subscription Analysis

Understanding Customer Behavior and Revenue Trends

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GitHub Repo: [View Here](#)

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INTRODUCTION

- **Goal:** To analyze streaming video subscription data to understand customer behavior, identify key drivers of churn, and provide actionable insights for business growth.
- **Dataset:** Subscription Cohort Analysis Data.csv, containing information on customer subscriptions, payment status, cancellation dates, etc.
- **Methodology:** Data cleaning, exploratory data analysis, customer segmentation, and churn prediction using a machine learning model.

IMPORT LIBRARIES

Started by importing essential libraries for data manipulation, analysis, and visualization:

- **Pandas:** For data handling and manipulation using DataFrames.
- **NumPy:** For numerical computations and array operations.
- **Matplotlib:** For creating static, interactive, and animated visualizations.

DATA EXPLORATION

Used `subs_df.info()` to:

- Understand the structure of the dataset, including the number of rows and columns.
- Check data types of each column (e.g., integer, float, object).
- Identify missing values (null or NaN) in any columns.



`subs_df.info()`



<class 'pandas.core.frame.DataFrame'>

RangeIndex: 3069 entries, 0 to 3068

Data columns (total 6 columns):

#	Column	Non-Null Count	Dtype
0	customer_id	3069 non-null	int64
1	created_date	3069 non-null	object
2	canceled_date	2004 non-null	object
3	subscription_cost	3069 non-null	int64
4	subscription_interval	3069 non-null	object
5	was_subscription_paid	3069 non-null	object

dtypes: int64(2), object(4)

memory usage: 144.0+ KB

DATA TRANSFORMATION

	customer_id	created_date	canceled_date	was_subscription_paid
0	154536156	2022-09-01	NaN	1
1	149713408	2022-09-01	2022-09-02	0
2	153756284	2022-09-01	2022-09-02	0

Recoding Categorical Variables: Converted categorical variables like 'was_subscription_paid' (Yes/No) into numerical format (1/0) for memory efficiency and compatibility with machine learning models.

Date Handling: Transformed 'created_date' and 'canceled_date' columns into datetime objects for easier date-based calculations and analysis.

FEATURE ENGINEERING

- Created new columns, such as 'created_month' and 'canceled_month,' by extracting the month from the respective date columns. This enabled monthly analysis of churn and other metrics.
- Calculated the 'subscription_duration' by subtracting the creation date from the cancellation date, providing a measure of how long each subscription lasted.
- Categorized subscription durations into groups ('0-7 days', '8-30 days', '31-60 days', '61-90 days', '91-120 days', '121-150 days', '151-180 days', '181-365 days', '1-2 years') using the `pd.cut` function to analyze retention patterns across different duration ranges.

CHURN ANALYSIS

Category	Total	Subscription Paid	Subscription Not Paid
Total Subscriptions	3069 (100%)	2936 (95.66%)	133 (4.34%)
Cancellations	2004 (65.29%)	1881 (93.86%)	123 (6.14%)

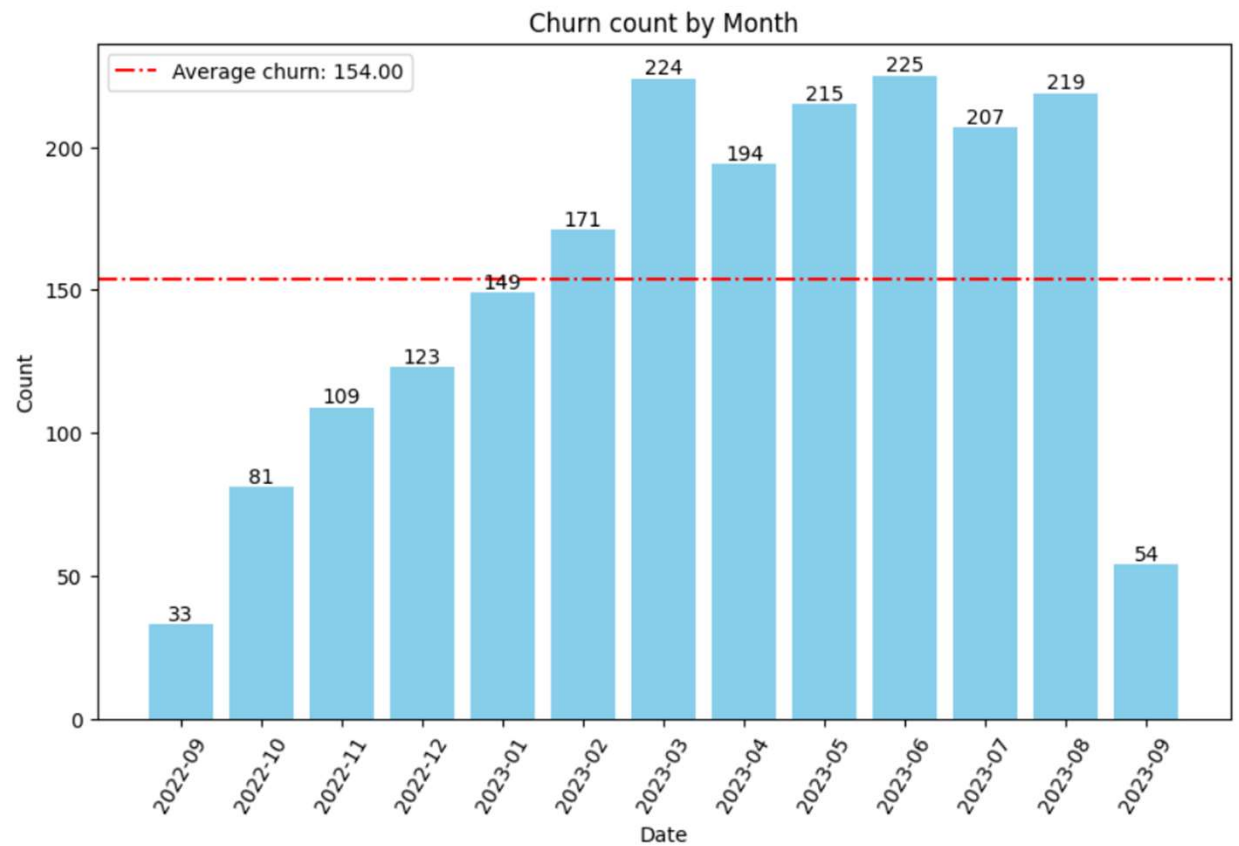
Higher Cancellation among Non-Paying Users indicates that free users are more likely to churn after their initial engagement.

Retention Strategy Opportunity:

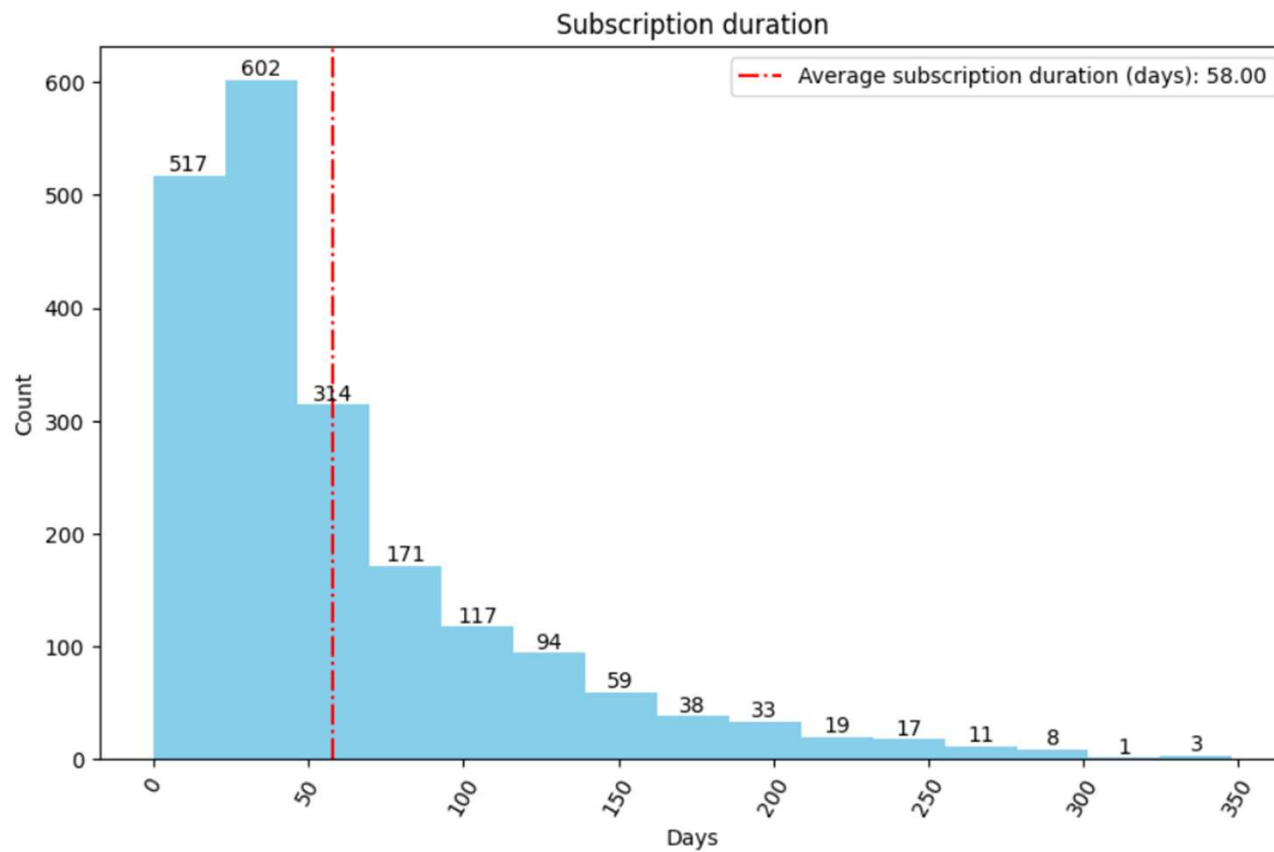
Paid users show a stable retention pattern, but non-paying users present an opportunity for improvement. Focusing on converting free users into paying subscribers can reduce cancellations and enhance long-term growth.

CHURN ANALYSIS

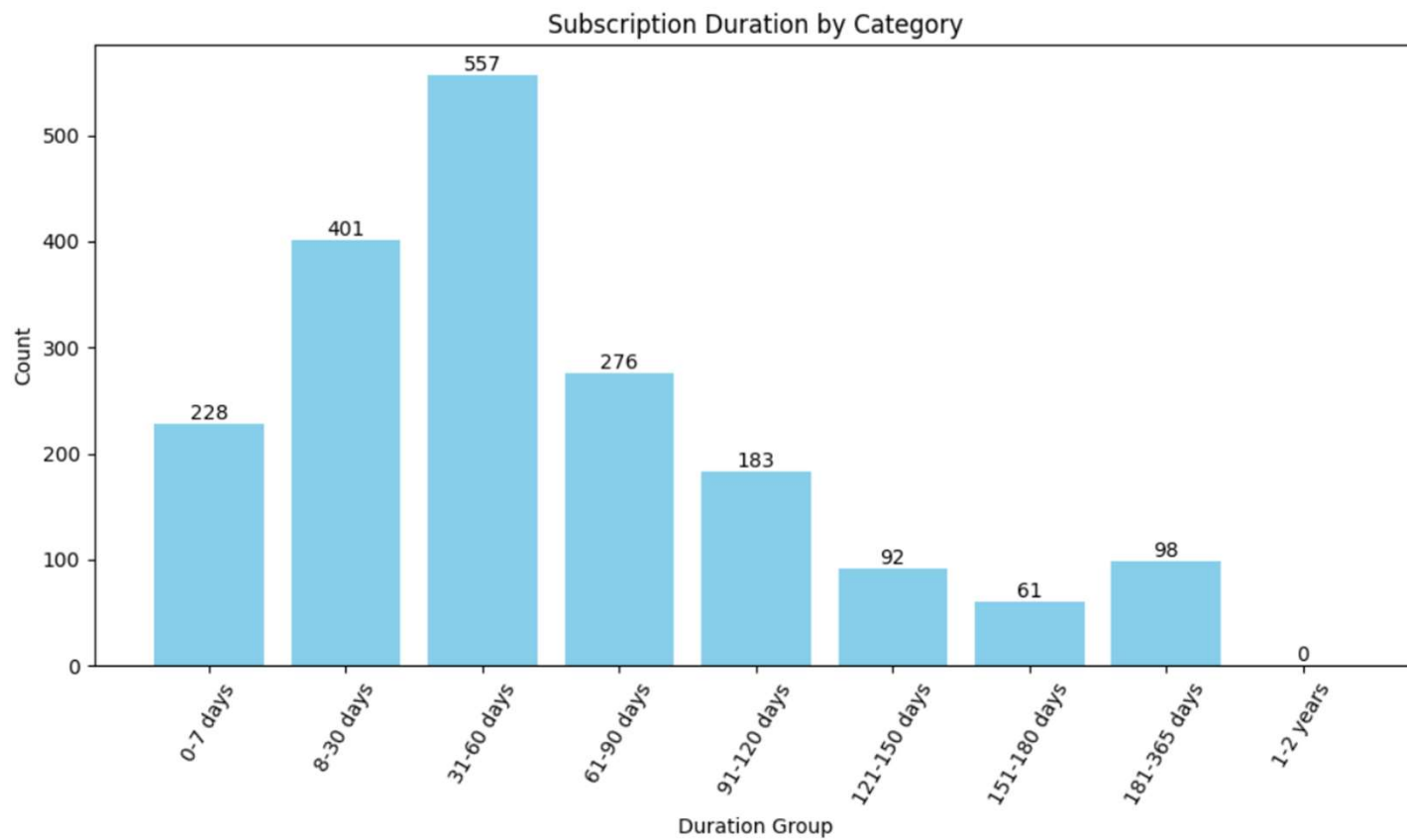
**Total
Customers
Cancelled their
Subscription
by Month**



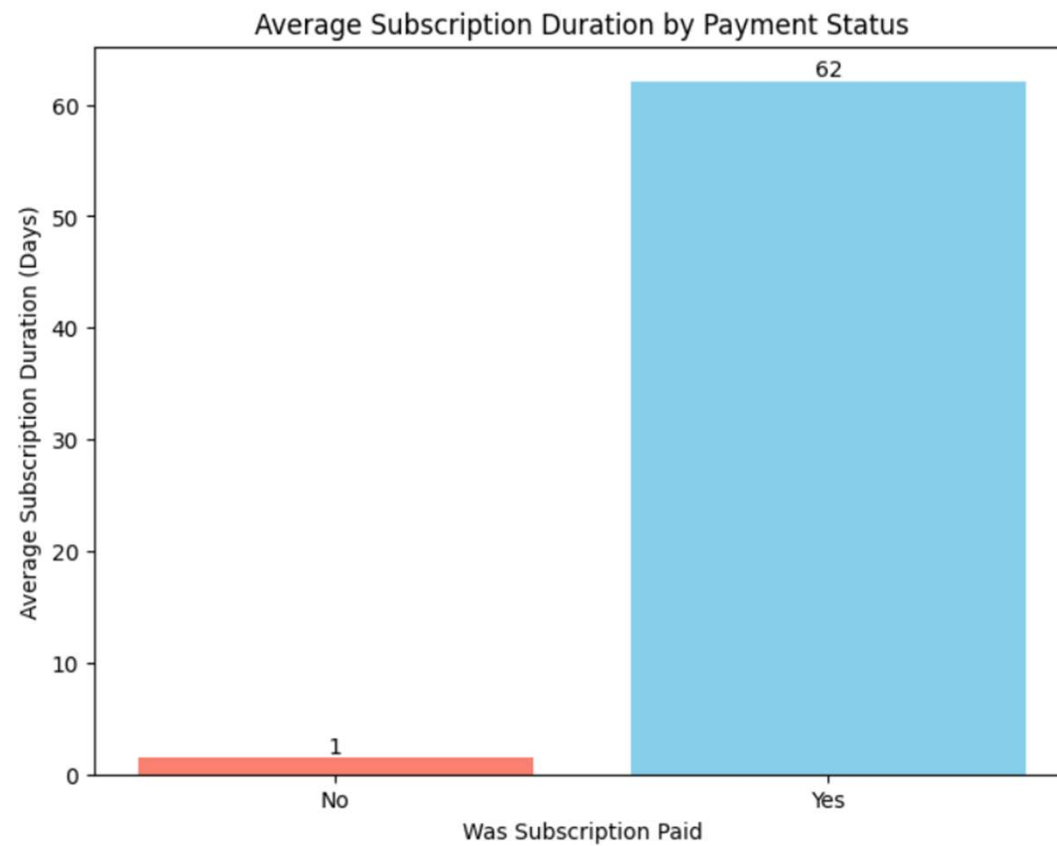
RETENTION ANALYSIS



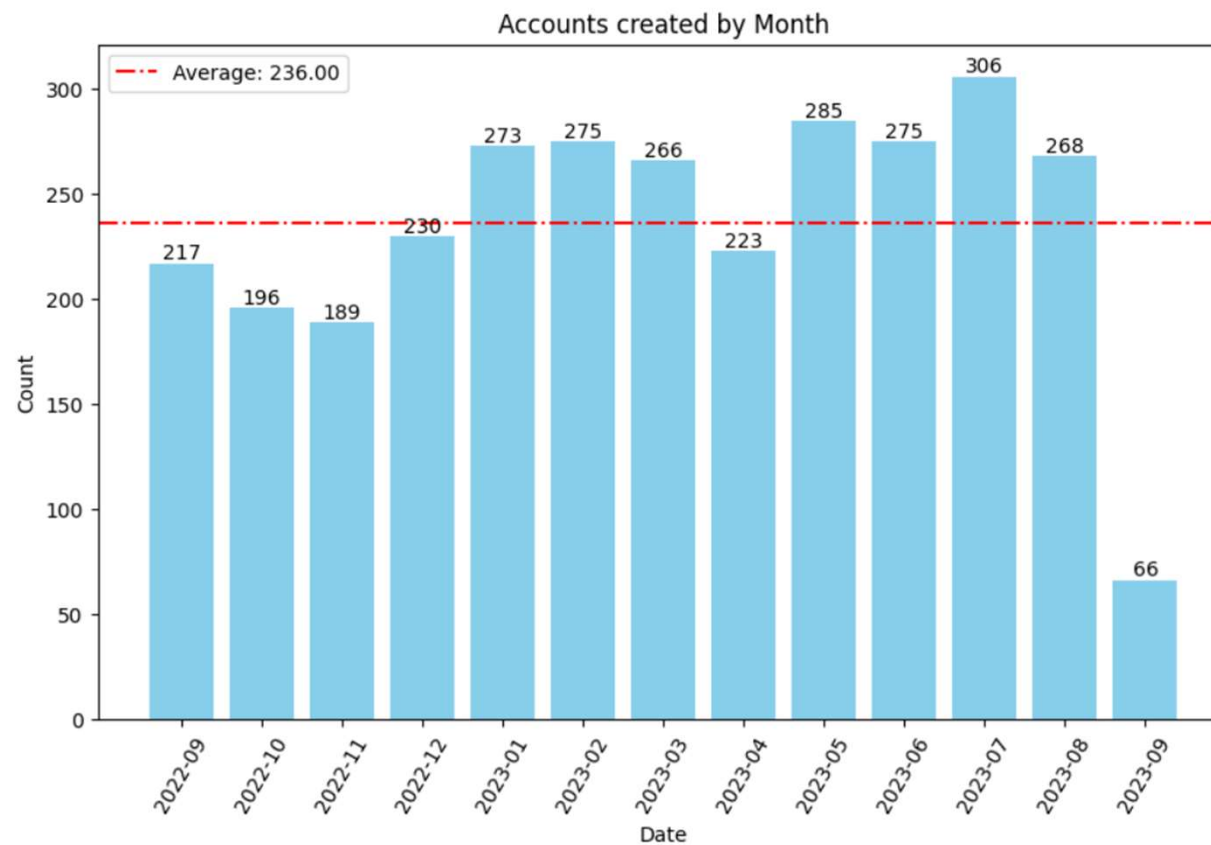
RETENTION ANALYSIS



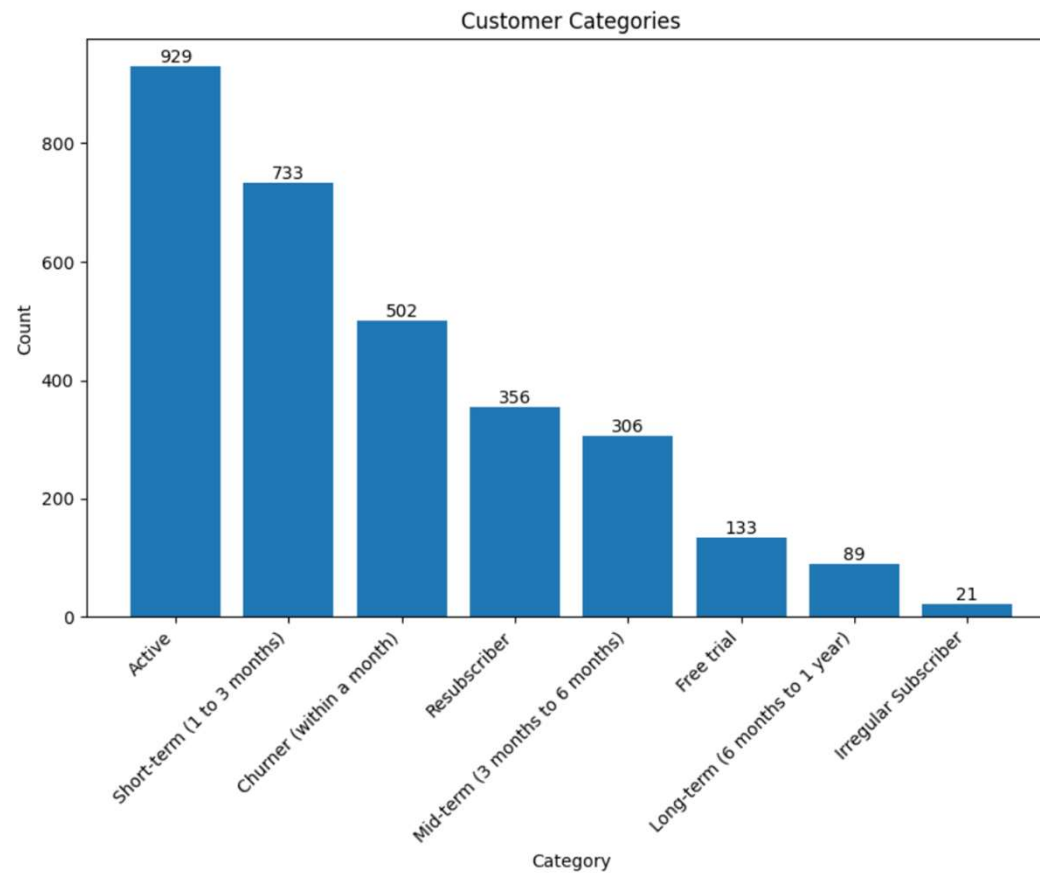
RETENTION ANALYSIS



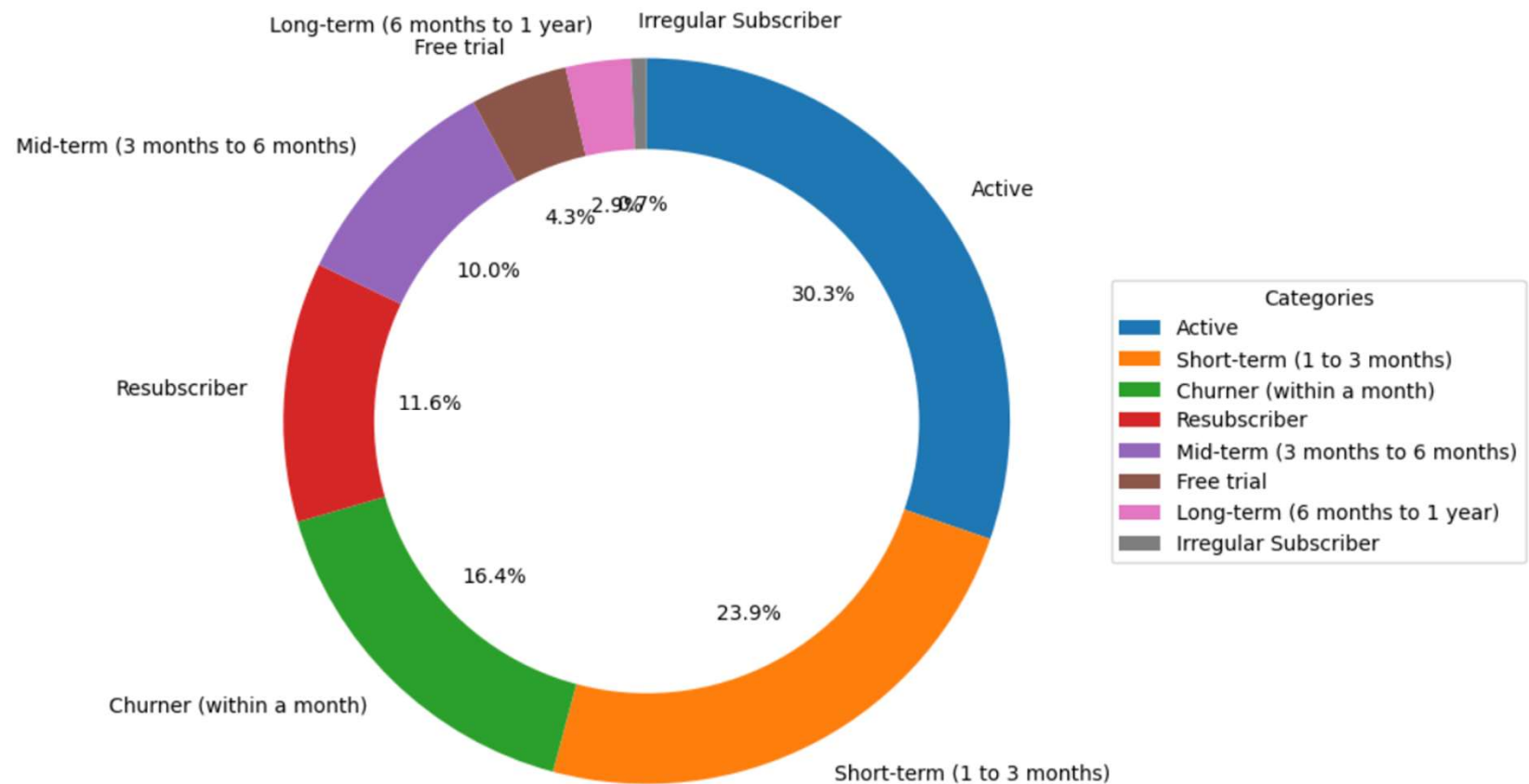
MONTHLY NEW SUBSCRIBERS



CUSTOMER SEGMENTATION



CUSTOMER SEGMENTATION



CONCLUSION



Key Insights: Identified key factors driving churn, including subscription duration, payment status, and customer categories.



Actionable Strategies: Developed data-driven recommendations for reducing churn, such as targeted interventions for high-risk segments and enhancing customer engagement.



Future Work: Outlined areas for future analysis, including investigating external factors and exploring more advanced predictive models.



Overall, this analysis provides a comprehensive understanding of subscription churn and empowers the business to make data-informed decisions for improved customer retention.

THANKYOU

Thank you for your time and attention throughout this presentation. I appreciate your interest and hope my work provided valuable insights.

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I look forward to connecting!