

## Executive Summary

### Customer Churn Decision Analysis

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#### 1. Problem Statement

The company is experiencing customer churn but lacks clarity on **why customers leave** and **which actions will meaningfully reduce churn**.

Due to limited responses from existing customers and the absence of fully aligned real-world datasets, this analysis uses a **synthetic but behaviorally realistic dataset** designed to mirror early-stage subscription businesses. The objective is not prediction accuracy for its own sake, but **decision support**: identifying actionable churn drivers and practical interventions.

#### Core question:

Why are customers leaving, and what should the company do next?

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#### 2. Key Findings

##### 2.1 Primary Drivers of Churn

The analysis consistently identifies four factors associated with higher churn risk:

1. **Customer complaints**

Customers who raise multiple complaints are significantly more likely to churn. Complaint volume is the strongest signal of dissatisfaction.

2. **Low customer tenure**

Customers in their first few months are at much higher risk of leaving, indicating weak early-stage engagement or onboarding.

3. **Low usage frequency**

Infrequent product usage correlates strongly with churn, suggesting customers who do not form habits disengage quickly.

4. **Heavy reliance on discounts**

Customers who require frequent or deep discounts tend to churn at higher rates, indicating price sensitivity rather than long-term value alignment.

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##### 2.2 What Does *Not* Drive Retention

- Simply offering more discounts does **not** improve long-term retention
  - Plan type alone (Free vs Paid) is not the primary churn driver
  - Monthly price differences matter less than perceived value and experience
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### **3. What the Company Should Do Next**

#### **3.1 Fix the First 90 Days**

The highest churn risk occurs early in the customer lifecycle.

**Recommended actions:**

- Improve onboarding clarity and guidance
- Proactively check in with new users
- Track early usage signals and intervene fast

Early retention will have the highest ROI.

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#### **3.2 Treat Complaints as Churn Alarms**

Complaints are not just service issues — they are **leading churn indicators**.

**Recommended actions:**

- Prioritize complaint resolution speed
  - Flag customers with multiple complaints for proactive outreach
  - Track complaint patterns, not just counts
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#### **3.3 Reduce Discount Dependency**

Discount-driven retention creates unstable customers.

**Recommended actions:**

- Shift from blanket discounts to value-based incentives
  - Use discounts selectively for onboarding, not long-term retention
  - Reinforce product value instead of price cuts
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#### **3.4 Increase Meaningful Engagement**

Retention improves when customers use the product regularly and purposefully.

**Recommended actions:**

- Identify features tied to repeated usage

- Nudge low-usage customers with targeted prompts
  - Measure success by engagement, not just logins
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#### 4. What the Company Should Not Do

- Do **not** chase churn reduction purely with promotions
  - Do **not** rely solely on predictive scores without interpretation
  - Do **not** over-invest in complex models before fixing fundamentals
  - Do **not** treat churn as a single metric instead of a customer journey
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#### 5. Risks & Limitations

- The dataset used is synthetic due to real-world data constraints
- Absolute churn rates should not be interpreted as exact forecasts
- Behavioral patterns are realistic, but not company-specific
- Results should guide **directional decisions**, not final policy

To mitigate these risks, the same framework should be applied once real customer data becomes available.

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#### 6. Final Recommendation

Churn is not primarily a modeling problem — it is a **product and experience problem**.

The most effective path forward is:

- Strengthen early customer experience
- Resolve dissatisfaction before it compounds
- Build engagement before optimizing pricing

This approach will reduce churn more sustainably than discounts or aggressive prediction systems.