

# Churn Analytics Retention - KPI Definition Doc

**Project:** Streaming Subscription Churn Analytics

**Company:** Drillinsight

**Version:** 1.0

**Date:** 2025-12-04

**Owner:** Drillinsight Data & Analytics Team

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## 1. Purpose

This document defines the **Key Performance Indicators (KPIs)** used in the Streaming Subscription Churn Analytics initiative.

It provides:

- Clear metric definitions
- Business rationale
- Calculation rules and formulas
- Edge cases
- Data dependencies
- Examples

These definitions serve as the **single source of truth** for Product, Growth, Finance, and Data teams.

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## 2. KPI Framework Overview

The KPIs fall into five major categories:

1. **Subscription Metrics**
2. **Churn & Retention Metrics**
3. **Cohort Metrics**
4. **Engagement & Behavior Metrics**

### 3. Definitions

#### 3.1 Subscription Metrics

##### 3.1.1 Active Subscribers

**Definition:**

Number of users with an active subscription at a given point in time.

**Formula:**

Code block

```
1 Active Subscribers = COUNT(users WHERE status = 'active')
```

**Time Grain:** Daily, Weekly, Monthly

**Use Cases:**

- MRR forecasting
- Active base monitoring
- Seasonal patterns

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##### 3.1.2 Monthly Recurring Revenue (MRR)

**Definition:**

Total subscription revenue expected from all active subscribers for a given month.

**Formula:**

Code block

```
1 MRR = SUM(price for all users WHERE status = 'active')
```

**Notes:**

- Does not include one-time promos
  - Uses subscription plan price, not total\_charges
-

### 3.1.3 Average Revenue Per User (ARPU)

#### Definition:

Average expected monthly revenue per active user.

#### Formula:

Code block

```
1 ARPU = MRR / Active Subscribers
```

#### Use Cases:

- Plan optimization
  - Revenue segmentation
- 

## 3.2 Churn & Retention Metrics

### 3.2.1 Churned User

#### Definition:

A user whose subscription has ended (cancel\_date is not NULL).

#### Business Rule:

Use `cancel_date` for month attribution:

- Cancelled on **2024-03-15** → counts as March churn
- 

### 3.2.2 Monthly Churn Rate (Primary KPI)

#### Official Drillinsight Definition

Code block

```
1 Monthly Churn Rate = (# of churned users in month N) / (active subscribers at start of N)
```

#### Example:

- Active on March 1: 7,000 users
  - Cancelled in March: 350 users
-

```
1 Churn Rate = 350 / 7000 = 5.0%
```

### Rationale:

This definition aligns with SaaS industry standards and Finance expectations.

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## 3.2.3 Retention Rate (Monthly)

Code block

```
1 Monthly Retention Rate = 1 - Monthly Churn Rate
```

Or explicitly:

Code block

```
1 Retention Rate =  
2 (# users active at end of month N) / (# users active at start of month N)
```

Use Cases:

- Month-over-month retention tracking
  - Comparing plans/segments
- 

## 3.2.4 Customer Lifetime Value (CLV)

Formula:

Code block

```
1 CLV = ARPU × Average Customer Lifetime (months)
```

If lifetime = 1 / churn rate:

Code block

```
1 CLV ≈ ARPU / Churn Rate
```

Use Cases:

- Acquisition budget decisions
  - Segment profitability
- 

## 3.3 Cohort Metrics

Cohort = group of users grouped by **signup month** (default).

### 3.3.1 Monthly Cohort Size

Code block

```
1 Cohort Size = COUNT(users WHERE signup_date in cohort_month)
```

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### 3.3.2 Cohort Retention Rate (Month N)

#### Definition:

Percentage of a cohort still active N months after signup.

#### Formula:

Code block

```
1 Cohort Retention (Month N) =  
2 (# cohort users active in Month N) / (cohort size)
```

#### Example:

Signup Cohort = 1,000

Active in Month 3 = 720

Code block

```
1 Retention = 720 / 1000 = 72%
```

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### 3.3.3 Cohort Churn Rate (Month N)

Code block

```
1 Cohort Churn (Month N) =  
2 1 - Cohort Retention (Month N)
```

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### 3.3.4 Rolling Retention

#### Definition:

User is considered “retained” if they are active **at any time after month N**, not necessarily every month.

Used in gaming & streaming.

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## 3.4 Engagement & Behavior Metrics

Behavior comes from `user_events`.

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### 3.4.1 Daily Active Users (DAU)

Users with  $\geq 1$  event on a given day.

Code block

```
1 DAU = COUNT(DISTINCT user_id WHERE DATE(event_time)=date)
```

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### 3.4.2 Weekly Active Users (WAU)

Users with  $\geq 1$  event in the last 7 days.

Code block

```
1 WAU = COUNT(DISTINCT user_id WHERE event_time >= today-7)
```

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### 3.4.3 Monthly Active Users (MAU)

Same logic but 30 days.

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### 3.4.4 Engagement Score (Composite Metric)

**Purpose:** Quantify user activity level.

Suggested formula:

Code block

```
1 Engagement Score =  
2 1 × (logins last 7d) +  
3 2 × (watch events last 7d) +  
4 1 × (clicks last 7d)
```

We weight **watch events** higher because they directly reflect product value.

Use Cases:

- Segmenting “low engagement → high churn risk” users
- Prioritizing retention campaigns

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### 3.4.5 Silent Users (Critical Churn Indicator)

**Definition:**

Active subscribers with **zero events in last 60 days**.

Code block

```
1 Silent User =  
2 status = 'active'  
3 AND no events in past 60 days
```

**Why important:**

Silent users typically have **4–7× higher churn** probability.

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### 3.4.6 Behavior Drop (30-day / 7-day)

**Definition:**

Percentage drop in engagement relative to previous period.

Code block

```
1 30d Drop =  
2 (events in last 30d - events in previous 30d)  
3 / (events in previous 30d)
```

Users with >50% drop in 30 days or >80% drop in 7 days are flagged as **early churn signals** (industry standard).

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### 3.5 Funnel Metrics

Funnel stages for streaming subscription service:

- 1. Signup
- 2. First Login
- 3. First Watch
- 4. Engagement (3 sessions within first week)
- 5. M30 Retention (still active after 30 days)

#### 3.5.1 Funnel Conversion Rate

Code block

```
1 Conversion (Stage A → Stage B)
2 = (# reached Stage B) / (# reached Stage A)
```

**Example:**

Signup → First Watch

- Signup users: 1,000
- Watch at least one video: 750

Code block

```
1 Conversion = 750 / 1000 = 75%
```

Use Cases:

- Onboarding optimization
  - Identifying friction points
- 

### 3.6 Model Performance Metrics

#### 3.6.1 Recall (Primary Model KPI)

Reason:



In churn prediction, **catching churners** matters more than avoiding false alarms.

Code block

```
1 Recall = True Positives / (True Positives + False Negatives)
```

Target:

- **Recall  $\geq 0.75$**  on top 20% predicted at-risk users.
- 

### 3.6.2 Precision

Code block

```
1 Precision = True Positives / (Predicted Positives)
```

Used to balance outreach cost.

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### 3.6.3 AUC (Area Under ROC Curve)

Measures overall model discrimination ability.

Target:

- **AUC  $\geq 0.75$**  acceptable
  - **AUC  $\geq 0.80$**  strong for churn
- 

### 3.6.4 Top-decile Lift

Code block

```
1 Lift =  
2 (Churn rate of top 10% risk group)  
3 / (Overall churn rate)
```

Target:

- **Lift  $\geq 3\times$**  indicates model usefulness
-

## 4. KPI Ownership

KPI Category	Owner	Team
Subscription metrics	Finance	Revenue
Churn & retention	Product + Finance	Growth Analytics
Cohort metrics	Product	Analytics
Behavior metrics	Product Intelligence	Data Analytics
Model metrics	Data Science	Data Analytics

## 5. Edge Cases & Exceptions

- Users with signup\_date AND cancel\_date in same month → count as churn in that month.
- Users with zero price plans (promos) → included but flagged.
- Users with event\_time out of range → excluded from engagement metrics.
- Fake users (no subscriptions) are excluded from churn KPIs.
- Churn date must be ≥ signup\_date; if not, flag in data quality check.

## 6. Data Sources

KPI	Table	Fields
Churn/ Retention	subscriptions	signup_date, cancel_date, status
Cohorts	subscriptions	signup_date
Engagement	user_events	event_time, event_type
Segmentation	users, subscriptions	demographics, add-ons
Pricing / Revenue	subscriptions	price, plan

# 7. Version History

Version	Date	Notes
1	12/4/2025	Initial KPI definitions