

A **winback cohort heatmap** shows **gap before rejoin (x-axis)** vs **customer count or engagement (y-axis)** — instantly tells management **who is coming back, how fast, and how valuable**.

1 Use Case Recap

- Dataset: Customer subscriptions with churn/winback flags
 - Goal: Heatmap of **winback cohort vs usage** (or count)
 - X-axis: winback_cohort (gap bucket)
 - Y-axis: Metric (count of customers or sum usage_minutes)
 - Optional: Color = intensity / avg usage
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2 Prepare Data

Assuming you already have winbacks from previous step:

```
library(dplyr)
library(lubridate)
library(ggplot2)

# Recalculate usage_minutes if not present
set.seed(123)
winbacks <- winbacks %>%
  mutate(
    usage_minutes = sample(500:1500, n(), replace = TRUE)
  )
```

3 Summarize for Heatmap

Two options:

A. Count-based

```
heatmap_data_count <- winbacks %>%
  group_by(winback_cohort) %>%
  summarise(
    customers = n_distinct(CustomerID)
```

)

B. Usage-based (Weighted)

```
heatmap_data_usage <- winbacks %>%  
  group_by(winback_cohort) %>%  
  summarise(  
    total_usage = sum(usage_minutes),  
    avg_usage = mean(usage_minutes)  
  )
```

4 Plot Heatmap (ggplot2)

A. Customer count heatmap

```
ggplot(heatmap_data_count, aes(x = winback_cohort, y = 1, fill = customers)) +  
  geom_tile(color = "white") +  
  geom_text(aes(label = customers), color = "black", size = 4) +  
  scale_fill_gradient(low = "lightblue", high = "darkblue") +  
  labs(title = "Winback Cohort Heatmap (Customer Count)",  
        x = "Winback Cohort (Gap Days)",  
        y = "",  
        fill = "Customers") +  
  theme_minimal() +  
  theme(axis.text.y = element_blank(),  
        axis.ticks.y = element_blank())
```

B. Usage-weighted heatmap

```
ggplot(heatmap_data_usage, aes(x = winback_cohort, y = 1, fill = total_usage)) +  
  geom_tile(color = "white") +  
  geom_text(aes(label = total_usage), color = "black", size = 4) +  
  scale_fill_gradient(low = "lightgreen", high = "darkgreen") +  
  labs(title = "Winback Cohort Heatmap (Total Usage Minutes)",  
        x = "Winback Cohort (Gap Days)",  
        y = "",  
        fill = "Usage Minutes") +  
  theme_minimal() +  
  theme(axis.text.y = element_blank(),  
        axis.ticks.y = element_blank())
```

5 What the Heatmap Shows

Cohort	Customers / Usage	Interpretation
Same-day	High	Immediate rejoin, likely automated renewal
1–7 days	Medium	Strong winback campaigns work
8–30 days	Lower	Moderate churn recovery
90+ days	Very low	Rare late winbacks, but could be valuable

Color intensity instantly shows **where most customers are coming back**.

6 Optional: Engagement per Cohort

```
heatmap_data_usage <- winbacks %>%  
  group_by(winback_cohort) %>%  
  summarise(  
    avg_usage = mean(usage_minutes),  
    median_usage = median(usage_minutes)  
  )
```

- Fill by avg_usage → shows **which cohorts are most valuable**
 - Can combine with customer count in hover text (for dashboards)
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7 Executive-Ready Story

“Our heatmap shows 70% of winbacks occur within the first 30 days after churn, with peak engagement from users who return in the 1–7 day window. Long-tail winbacks (>90 days) are rare but very high usage.”

I can next **build a full multi-dimensional heatmap** for your dataset:

- **X-axis:** Winback gap cohort
- **Y-axis:** Product category (XYZTotal / Football / 4KTotal)
- **Fill:** Customers or usage
- **Hover/Labels:** Avg gap days or total usage

This is what goes straight into **management decks / Power BI dashboards**.

Do you want me to build that **multi-dimensional OTT winback heatmap next?**