

# **Cart Abandonment: Key Drivers and Actionable Insights**

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# Agenda

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- 05 **Our Suggestions** Slide 13
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# Business Problem

## What's the Problem?

- Items added, no order
- Missed expected orders
- Incomplete digital journeys

## Why Matters?

- Lost revenue
- Poor forecasting
- Delivery disruption
- Low digital engagement

## What This Means?

- Unstable order cycles
- Operational inefficiency
- Lower platform adoption
- Higher customer friction

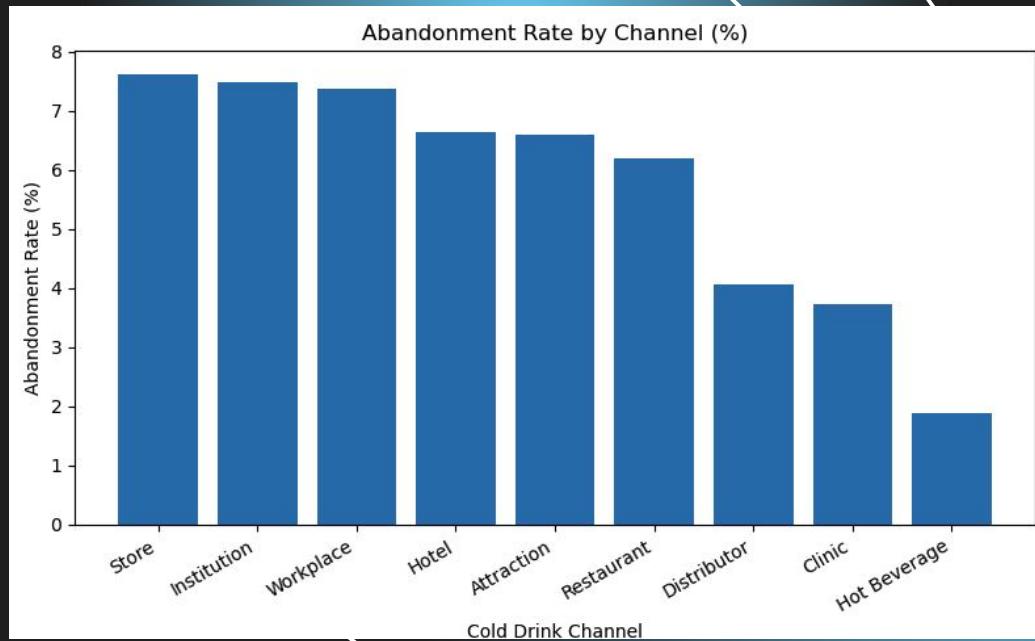
# Behavior Insights

## Device Patterns

- Tablet highest rate (8.41%)
- Mobile elevated rate (7.48%)
- Desktop most reliable (5.59%)

## Channel Patterns

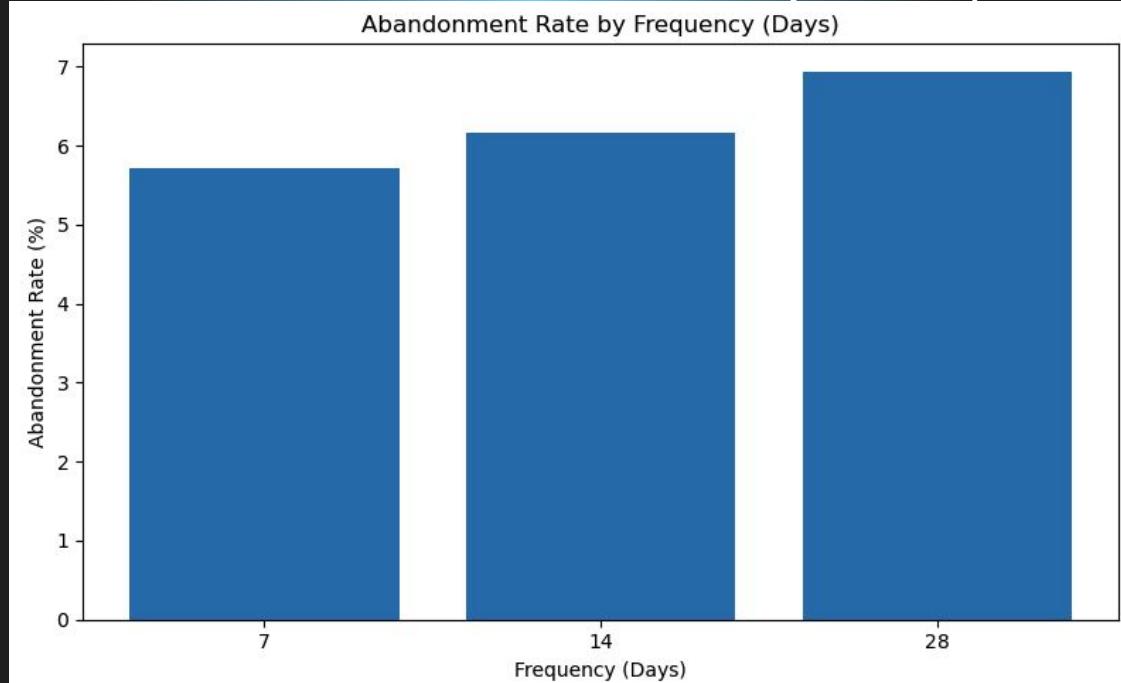
- Store
- Institution
- Workplace



# Customer Insights

## Order frequency pattern

- 28-day: 6.94% (highest)



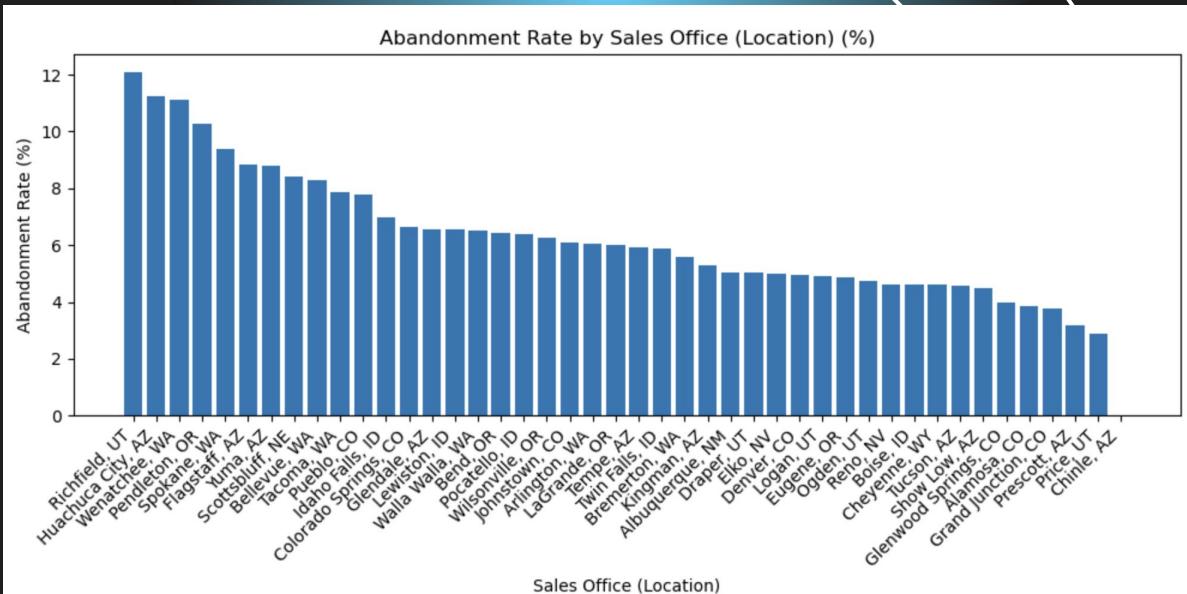
# Product and Office Insights

## Product Patterns

- Large packs drop more
- Manufacturing packs appear
- Special Flavors

## Location Patterns

- Clear high-risk offices
- Some >10–12% abandon
- Regional variation exists



# Key Behavioral Insights

## Behaviors Before Abandonment

- No checkout started
- Many product views
- Add → no follow-up
- Remove-from-cart actions

## What Happens After Abandonment

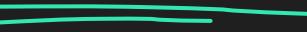
- Orders placed through other channels
- Order delayed to next cycle
- No activity for multiple windows

# Window Level Model

Predicts abandonment using aggregated clickstream data, running total sales and customer data per order window

Model	Accuracy	Recall_1	F-1_Score_1	AUC
Boosted Logistic Regression	0.93	0.87	0.71	0.95
Decision Tree	0.93	0.87	0.71	0.94
Logistic Regression	0.93	0.79	0.69	0.92
LightGBM	0.92	0.87	0.7	0.9
XGBoost	0.92	0.87	0.7	0.95

# Window Level Model



## Top factors increasing abandonment

begin\_checkout\_flag\_0

viewed\_product\_details\_1

mycoke\_cart\_count

## Top factors decreasing abandonment

begin\_checkout\_flag\_1

view\_product\_details\_0

total\_orders\_running

unique\_page\_views

# Customer Level Model

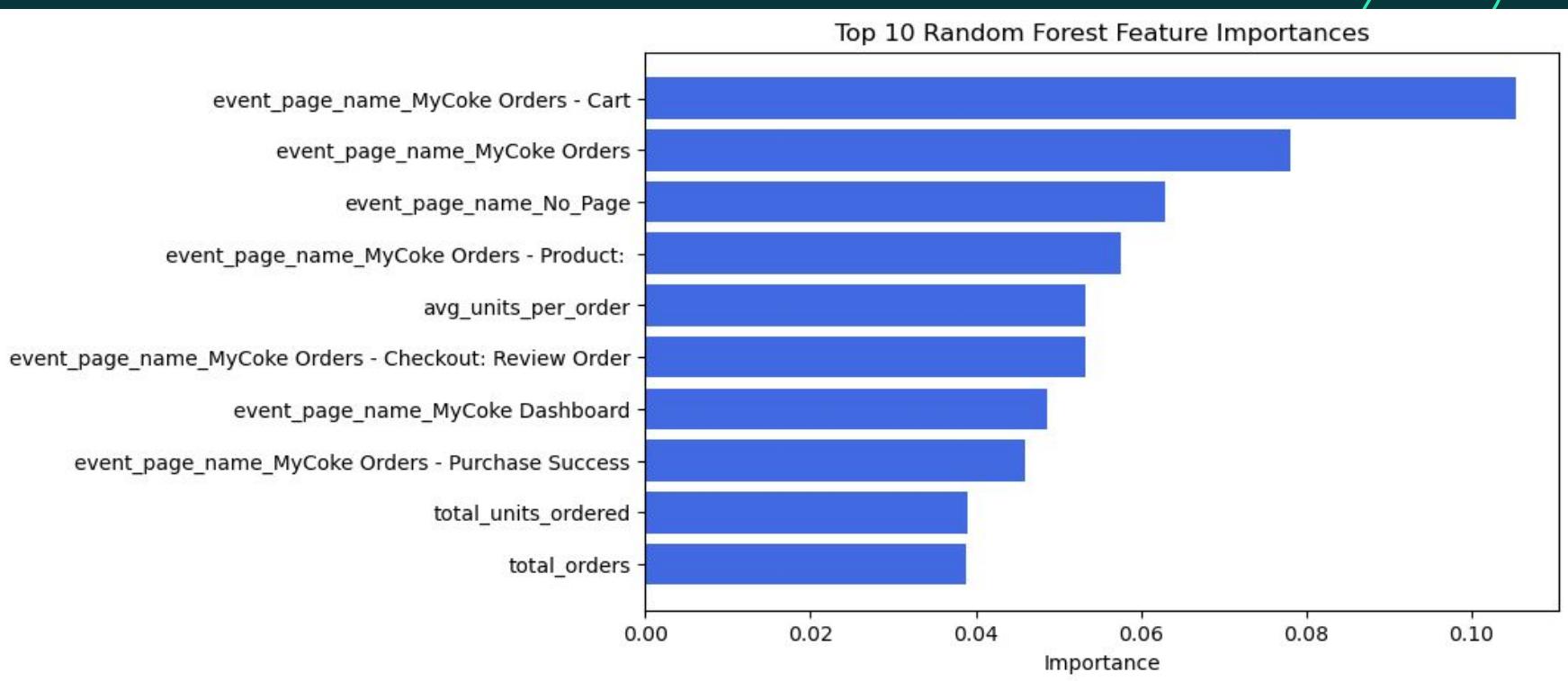
## Metrics

Predicts abandonment from consolidated client data

Outcome	Precision	Recall_1	F-1_Score_1
Abandon	0.79	0.48	0.58
Not Abandon	0.81	0.92	0.86
AUC			
0.868			

# Customer Level Analysis

## Top 10



# Customer Level Analysis

## Data Lost

### Customer

Windowed

Customers who did not engage with carts or website.

Google Analytics

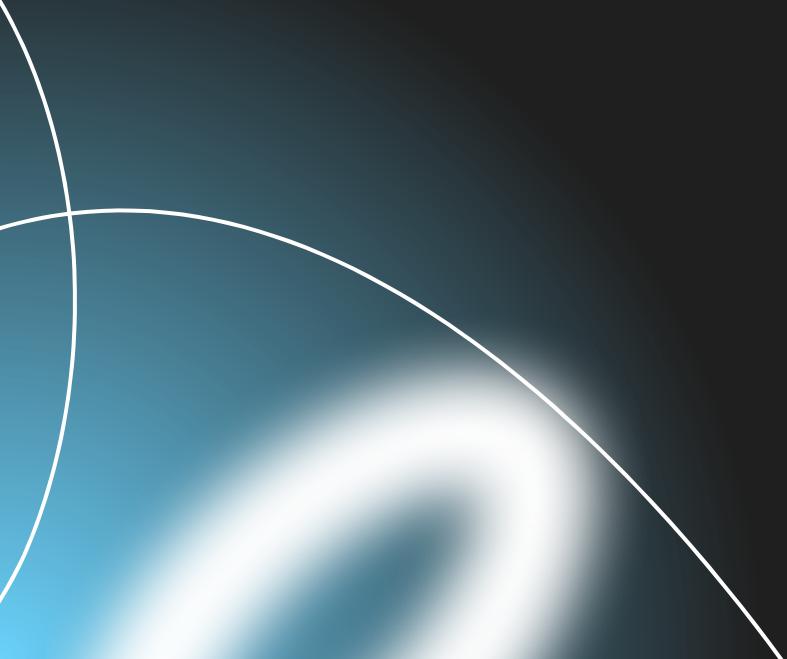
Customers where no orders were placed in the window.

Sales

Customers where no orders were placed in the window.

### Stagnant Customers

# Lessons Learned



1

The more a customer engages with the website, the more likely they are to abandon.

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2

Larger order sizes are less likely to be abandoned.

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3

Abandonment behavior repeats.

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# Financial Impact

## Window-Level Summary

- Abandoned order windows: **3,092**
- Estimated revenue per unit: \$26.42
- Total revenue lost: **\$246,944**
- Avg value at stake per abandoned order: **\$79.87**

## Customer-Level Summary

- Total customers: 6,334
- Unstable customers: **1,752**
- Avg value at risk (unstable): **\$134.38**

## Business Opportunity Scenarios (Model + Recovery)

- 10% recovery: \$21,484 potential savings
- 20% recovery: \$42,968 potential savings
- 25% recovery: \$53,710 potential savings
- 50% recovery: \$107,421 potential savings
- 75% recovery: \$161,131 potential savings
- 100% recovery: \$214,841 potential savings

# Suggestions

## Customer

Run Monthly

Target chronic abandoners

Seasonal Contract

Order scheduling automation with incentive

## Window

Run Weekly

Window Model for specific order windows

Recommendations based on similar orders in region

# Suggestions

## Chronic abandoners

Stable pricing for committed schedules

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Contract revision

## Heavy browsers with purchase

Consultation on best product mix based on alike customers from account manager

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Product/preset bundles recommendations on website

# Q&A

# Thank you