

# Forms Growth and Analysis

# Agenda

- I. Data Analysis and Presentation of Findings
- II. Business Intelligence & Dashboarding
- III. Marketing

# Data Analysis and Presentation of Findings

# Process & Assumptions

## 1. Getting Started

- Imported data and cleaned it (handled missing values, split multi-state entries)
- Visualized basic stats to identify potential influential factors

## 2. Assumptions Made

- **Habit formation** is **sending 3 or more envelopes in the both of the last 2 months**
- Users with missing dates, companies, or multiple states are accurately represented
- The current features in the dataset are relevant for predicting habit formation

## 3. Pattern Identification

- Utilized logistic regression to assess feature importance
- Checked correlations to discern relationships
- Adjusted for class imbalance to refine model accuracy



# Results

- Model **excels at predicting no habit formed**, but needs tuning to adjust for **imbalanced data**
- Balancing (Oversampling minority)
  - Better identifies habits formed, but with some false positives
- Accuracy (91%)**
  - Decrease after oversampling due to precision, but still acceptable
- ROC-AUC Score: 0.9635**
  - strong ability of the model to distinguish between the two classes, regardless of the threshold set for classification
- Features**
  - States** - significant impact on habit adoption
  - User type** - moderate impact on habit adoption

Initial	precision	recall	f1-score	support
0 (no habit)	0.97	0.99	0.98	24981
1 (habit formed)	0.7	0.39	0.5	1346
accuracy			0.96	26326

Feature - State	Coefficient
South Carolina	1.35
Indiana	1.32
North Carolina	1.31
New Jersey	-1.41
Ontario	-1.51
New York	-1.76

Feature - User Type	Coefficient
Breeze User	1.07
Auditor	-0.91
Broker	-1.59
User Not Specified	-2.29

# Actionable Insights

## 1. Geographical Variation in Product Adoption

- States like South Carolina, Indiana, and North Carolina (US-SE) have the highest positive coefficients, indicating a strong positive correlation with habit formation
- On the flip side, states like New York, Ontario, and New Jersey (US-NE) show strong negative correlations
  - ➔ **Recommendation:** Investigate how marketing and sales approach different states, or how users differ across states (tech adoption, competitor usage rates, etc)

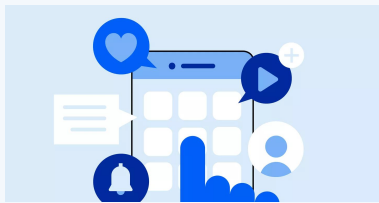
## 2. UserType Impact on Habit Formation

- Brokers and Auditors have strong negative correlations with habit formation
- On the contrary, the Breeze users have a positive correlation
  - ➔ **Recommendation:** Find out why brokers and auditors are less inclined to form habits. Figure out if they are using a different product, have less use cases, or are less likely to adopt technology

## 3. Interaction with Features and Habit Formation

- Certain product interactions, such as adding libraries, creating files, and adding forms, have a positive association with habit formation
  - ➔ **Recommendation:** Explore encouraging user interaction with these features to accelerate product habit adoption. Ideas include tutorials, in app suggestions, or customer marketing campaigns

# Additional Recommendations



## Dive into User Engagement Metrics

- Session duration
- Frequency of use
- Features used



## Assess External Factors

- Market analysis: account for local real estate market dynamics
- Competitors: determine if users in specific regions, or specific user types prefer a competitor



## User Profiling

- Demographics: age, years in real estate, education
- Tech proficiency
- User timeline

# Business Intelligence & Dashboarding