User Adoption Analysis – Report

Objective:

Identify which factors predict user adoption, defined as a user logging in on three separate days within a single seven-day period.

Below are the steps and methodology:

1. Data Understanding:

- takehome_users.csv: 12,000 user profiles including signup details and marketing status.
- **takehome_user_engagement.csv:** 208,000+ records of user logins with timestamps.

2. Labeling Adopted Users:

- A user was flagged as adopted if they logged in on 3 unique days within any rolling 7-day window.
- o Result: 1,656 users (13.8%) met this criteria.

3. Feature Engineering:

- Extracted features like:
 - days_since_creation
 - invited (binary: whether invited by another user)
 - One-hot encoded creation_source
- Converted session times and account creation times to datetime objects.

4. Modeling Approach:

Trained a Logistic Regression model to predict adoption.
Split data into train/test with stratification.

Findings:

| Feature | Impact on Adoption |
|----------------------------|---|
| opted_in_to_mailing_list | Strong positive predictor |
| enabled_for_marketing_drip | Negative correlation (may indicate passivity) |
| creation_source | Google Auth users showed higher adoption |
| days_since_creation | Longer account age correlated with adoption |

| invited | Slightly higher chance of adoption |
|---------|------------------------------------|
| | |

Model Performance:

Accuracy: 96.9%, F1-score (Adopted): 88.4%

Areas for Further Investigation:

- Analyze login frequency patterns and total login counts.
- Explore organization-level dynamics (e.g., adoption clusters within orgs).
- Incorporate time-to-first-login or activity bursts post-signup.

Conclusion:

Marketing opt-in and signup methods are significant predictors of user adoption. The presence of an inviter and prolonged engagement period also positively influences adoption likelihood.