Machine Learning Models:

Bust Out Fraud Detection

Vertical

- · Industry: Financial Services, Banking
- Industry Detail: Consumer Credit Card
- Function: Credit Card Issuer

Use Case Highlights

- Avoid credit loss by detecting those users who intend to rack up charges on a credit card but never pay.
- Collection of models using transactional history, payments, and non-monetary activity to detect and prevent fraud before it happens.
- Get more lead-time to take pre-emptive action by identifying bust-out customers before they occur with additional data sources, advanced modeling techniques, and decades of combined fraud experience.

Use Case

- A bust-out is a type of credit card fraud where an individual applies for a credit card, establishes a normal usage pattern
 and solid repayment history, then racks up numerous charges and maxes out the card with no intention of paying the
 hill
- The objective is to identify fraudsters early on before they rack up credit card charges.

Technical Highlights

- A non-linear adaptive analytics approach used for credit abuse detection to provide a better predictive power and ability to identify accounts earlier in the cycle.
- Dynamic Bust-out Pattern Analysis identifies interconnected patterns between transaction types and velocity, high risk purchases and non-monetary information.
 - Rapid: Quickly increase spend and make large payments to free up more opportunity to buy.
 - Collusive Merchants: Large purchases made at single (often newer) merchants.
 - Long Play: Spending and payments increase over many months. Customer continues to spend up to their full credit limit and then payments bounce.
- Neural Network Model for Bust-out Prediction significantly improves detection accuracy, 5 days earlier.
 - Neural Networks uncover the hidden non-linear relationships between inputs and outcomes to predict small probability
 events.
 - Customer activity patterns are monitored on a daily basis to identify predictive patterns.





Machine Learning Models:

Bust Out Fraud Detection

Business Impact

- \$60MM annual savings
- 5 to 7-day earlier identification
- No rise in false positives
- · Completely integrated into customer's environment

Data Sources and Features

- Efficiency of bust-out fraud detection increased by utilizing more data inputs and sophisticated neural network models to find more fraud.
- Developed an "After the Fact" model to accurately assess the real bust-out baseline, built predictive model using the expanded baseline, and finally estimated savings under various scenarios of analyst capacity.
- Transaction data examined to create variables based on high risk merchandise to capture more granular risk behaviors.

Leveraging Model Output

Data In

- · Early intervention
- · Card transactions
- Customer service interactions
- Underwriting
- · Transfer information
- · Payment information
- Customer demographics

Data Out

- See compromise window, pre-fraud distribution
- Probability score for Bust-out Fraud at account level
- · Score distribution over time
- · Card burn rate trends
- Card fraud usage distributions



