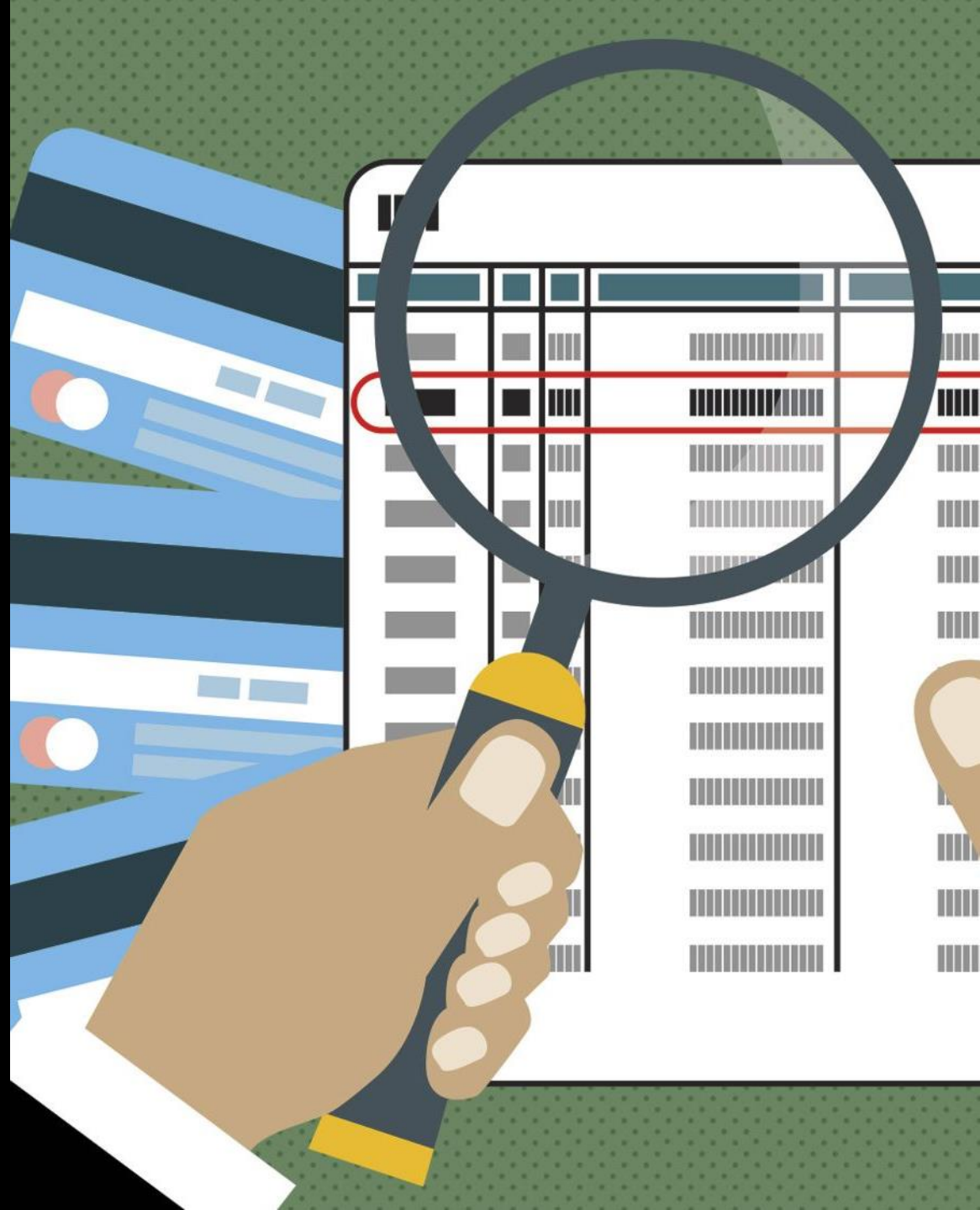


Credit Card Fraud Detection

Mai Tran

Metis

3/23/2022



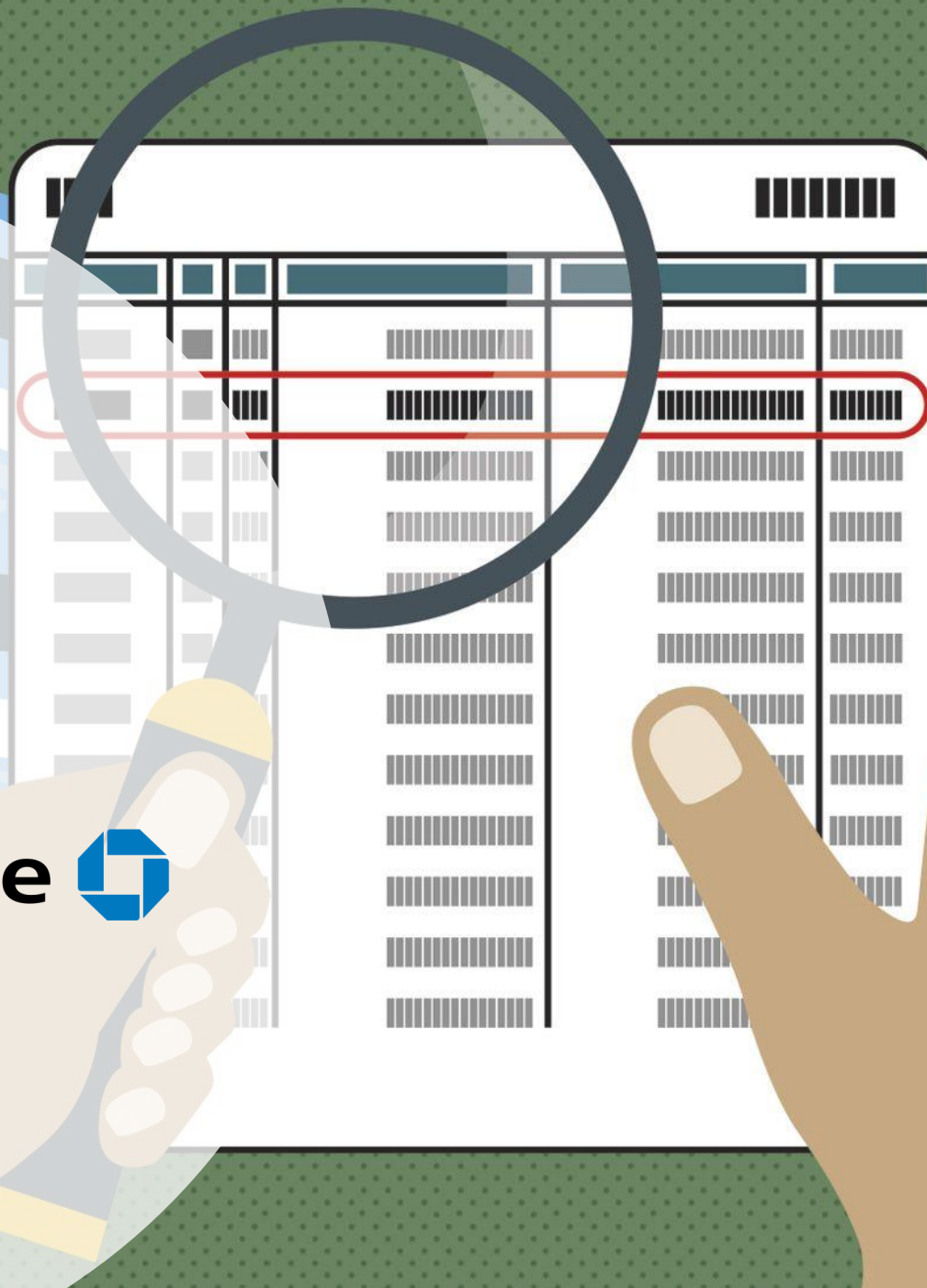
161%

of credit card fraud occurrences
have **increased between 2015 – 2020**, according to Atlas VPN.

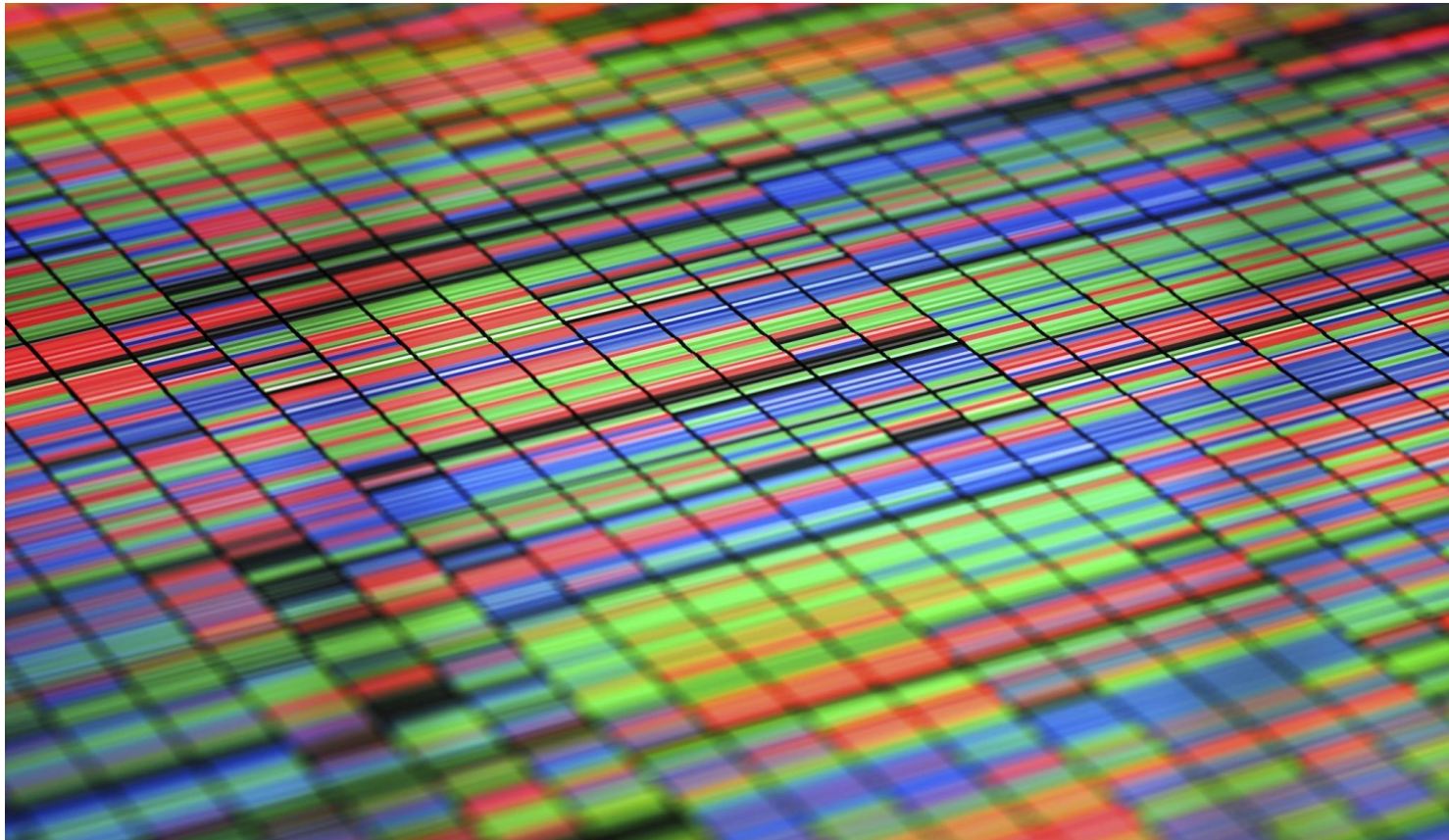


Client

JPMorganChase 



Data

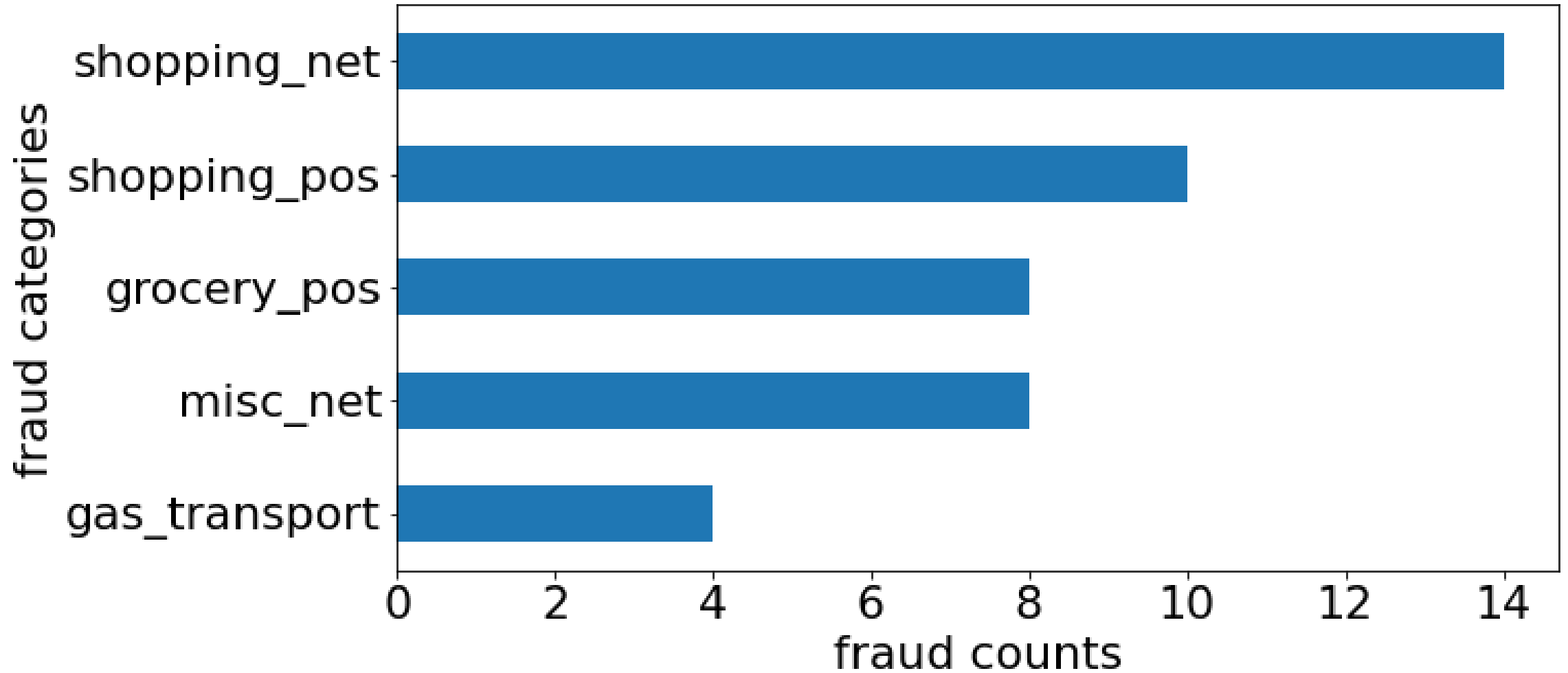


- Simulated data from Kaggle
- Highly imbalanced with 0.6% fraud rate
- 10,000 rows with 21 features

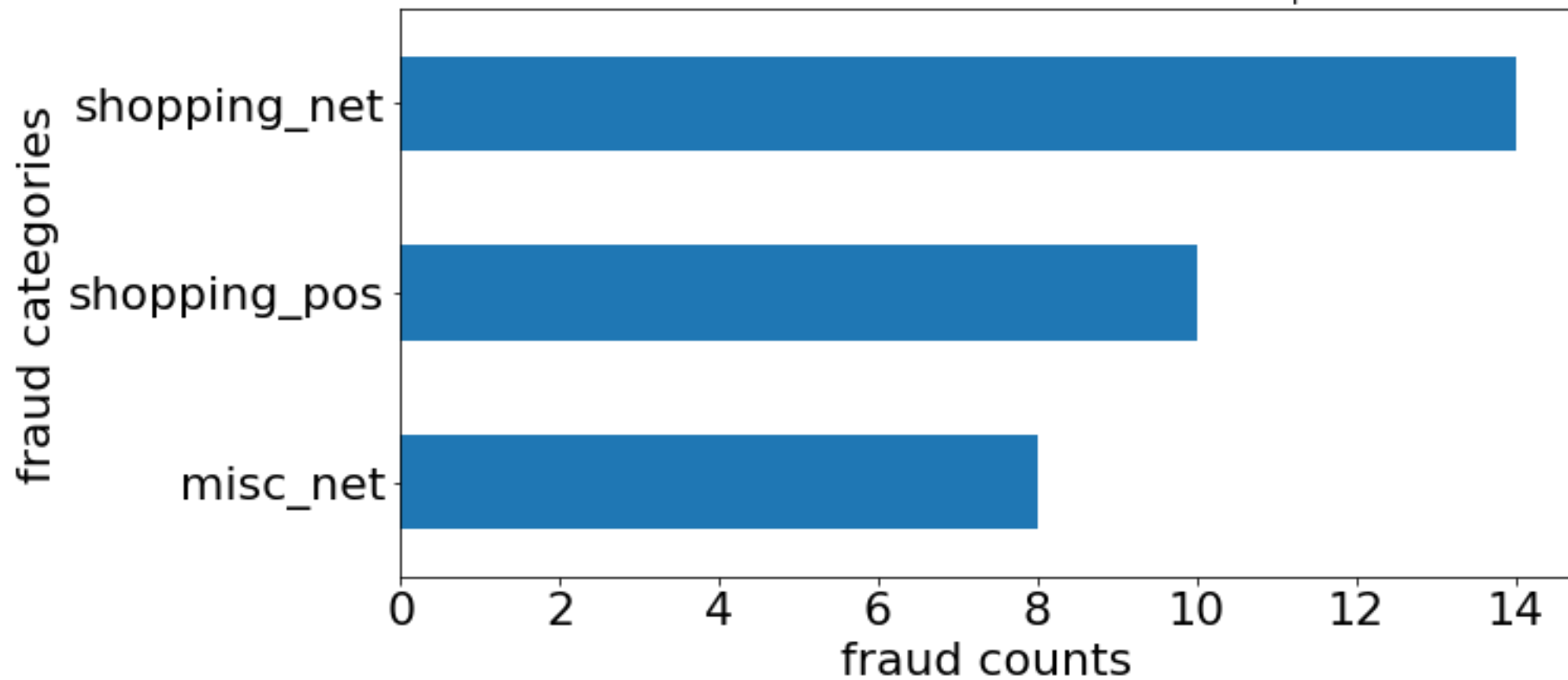
Fraudulent Transaction Statistics

Average	Median	Minimum	Maximum	Mode	Percent Fraud
\$558.71	\$719.50	\$6.90	\$1376.04	\$333.28	0.6 %

TOP 5 COMMON FRAUD CATEGORIES

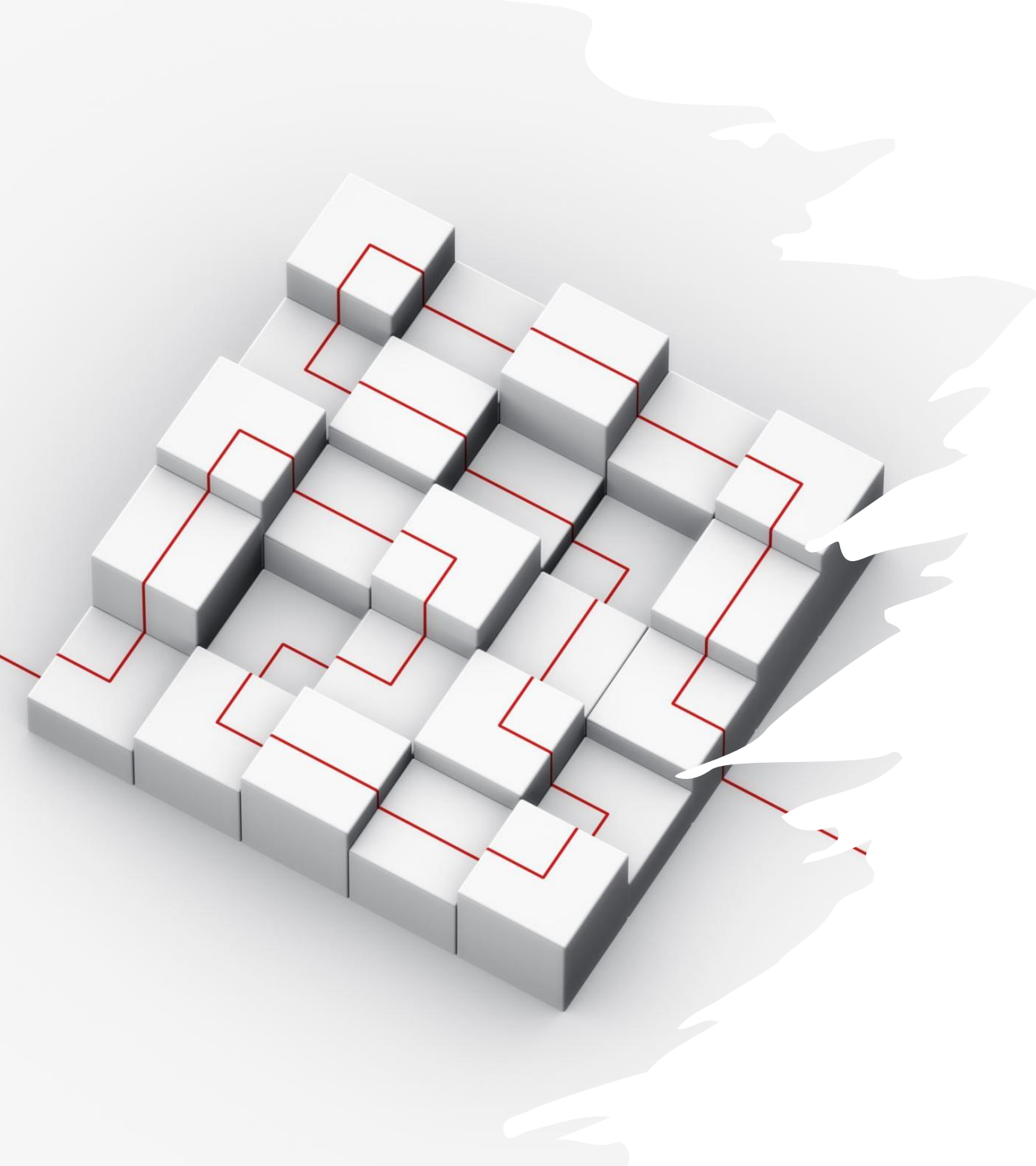


COMMON FRAUD CATEGORIES HIGHER THAN \$550 PER TRANSACTION



Project Design

1. Metric selection
2. Baseline Model – Logistic Regression
 - 1.Stratified sampling
 - 2.Class weight balancing
 - 3.Hyperparameter tuning
3. Compare baseline with 4 other models



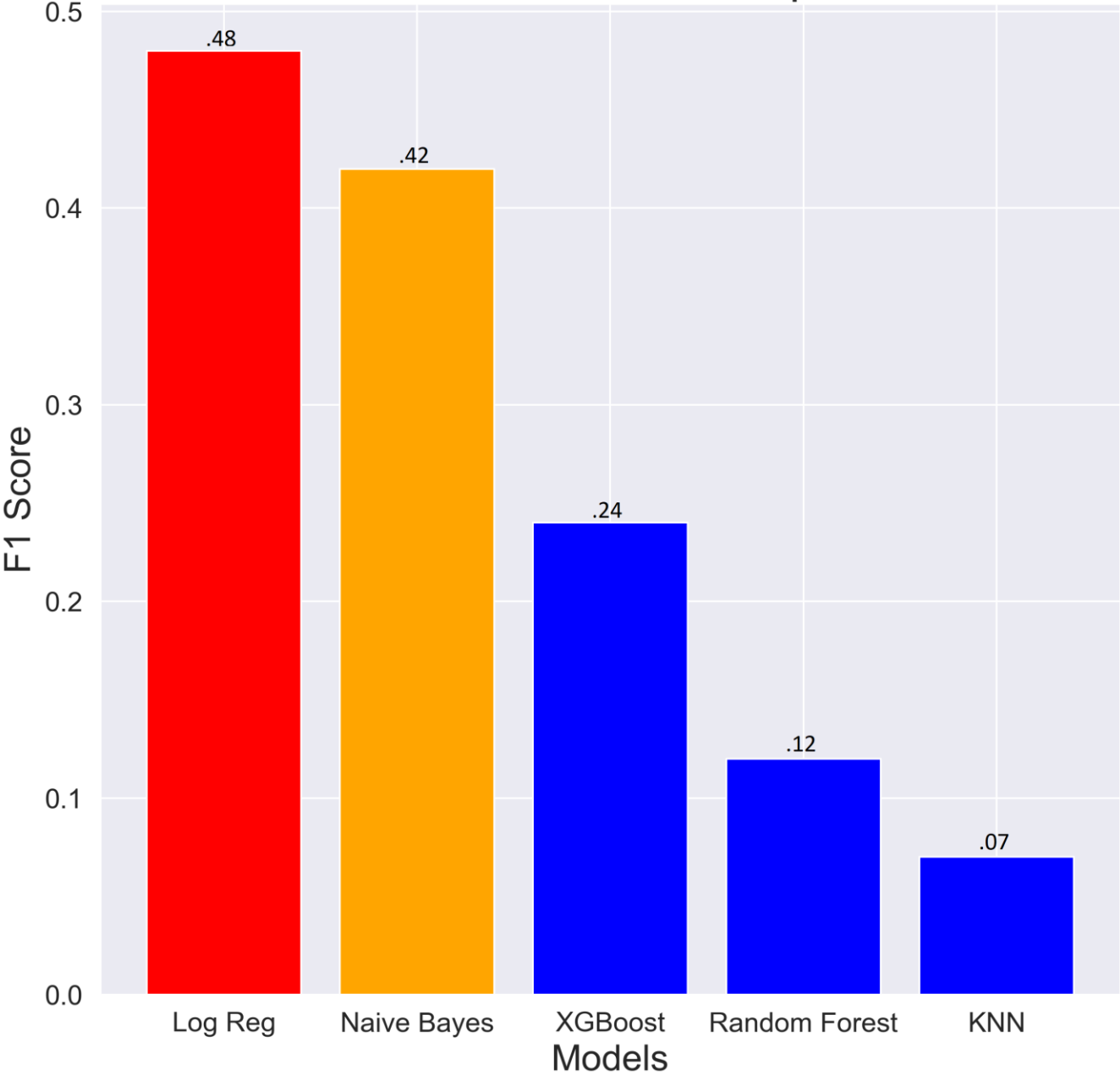
Metric

- F1 score was selected
- Vital financial assets – no compromise between recall and precision!
- High recall + high precision = high F1 score
- (fewer false negatives – those real malicious fraud charges that were identified as valid charges) + (fewer false positives – those falsely flagged fraud charges that were actually legitimate) = no one losing money + no one being annoyed and upset

Logistic Regression Tuning Results

Logistic Regression	F1 score
1) LR with class imbalance	0.0
2) LR with class_weight='balanced'	0.06
3) LR with best hyperparameters	0.18
4) LR with both best hyperparameters and class weights	0.47
5) LR with scaled features	0.48

Model Performance Comparison





Future Work

- Apply resampling techniques to reduce class imbalance: under-sampling, over-sampling, and SMOTE (synthetic minority oversampling technique)
- Integrate all 5 models into an ensemble classifier such as a voting ensemble or a stacked ensemble
- Time series analysis

Thank you!



Appendices

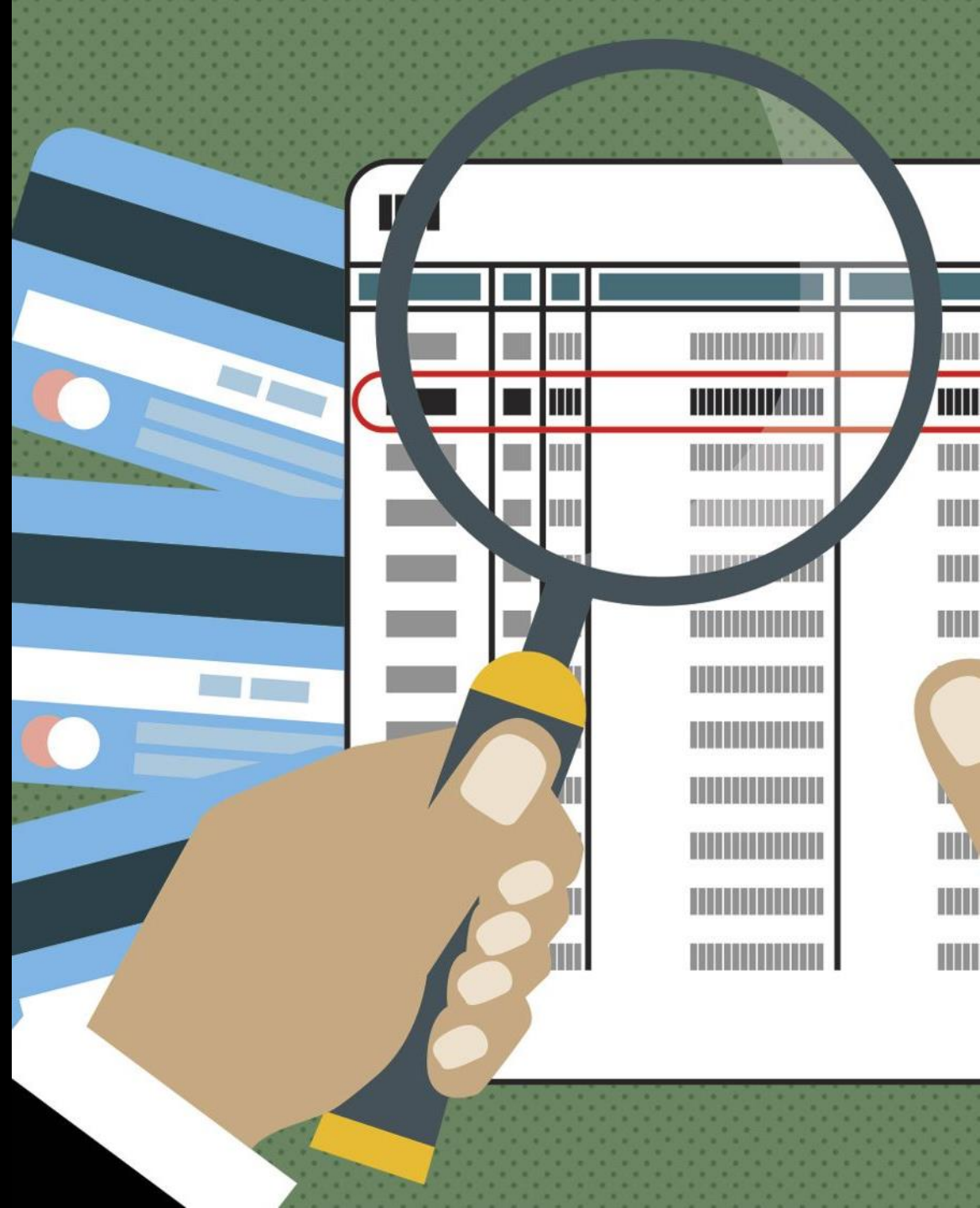
Biggest Incident in History

In **mid-2000s**, more than **32,000** credit card information was stolen resulting in **total financial losses** of 17 million Great British Pound (GBP) or **22 million US Dollars (USD)**.



Problems

- 1) how to catch more seemingly legitimate transactions that are actually legitimate?
- 2) how to catch more seemingly fraud transactions that are actually fraud?
- 3) Imbalanced data -- 0.6% fraud



Cost Benefit Analysis

- Cost – Positive Number
- Benefit – Negative Number

TN (Normal Transaction):	$C_{TP} = -1 \times \text{Transaction Amount} \times \text{Merchant Fee}$
FP (Falsely Flagged Fraud):	$C_{FP} = \text{Intervention Cost} - \text{Transaction Amount} \times \text{Merchant Fee} + \text{Customer Frustration}$
FN (Undetected Fraud):	$C_{FN} = \text{Transaction Amount}$
TP (Detected Fraud):	$C_{TP} = \text{Intervention Cost}$

Source: RCG Global Services