# Fraud Detection ML Pipeline Report

Comprehensive Analysis and Insights

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# **Executive Summary**

### TL;DR - Key Findings

Key Metrics	
Total Channels Analyzed	17,273
High-Risk Channels	2,624 (15.2%)
Channels with Anomalies	169 (1.0%)
Average Quality Score	3.44/10
Average Bot Rate	2.3%
Total Traffic Volume	1,189,903

#### **Critical Findings**

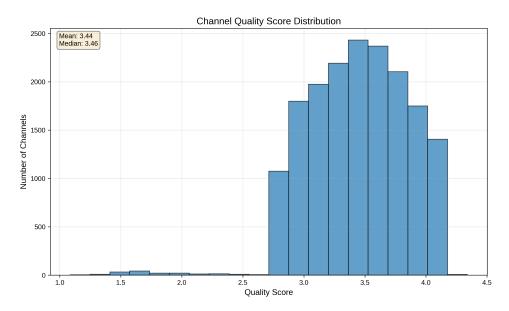
• High Risk: 2624 channels (15.2%) classified as high-risk

#### **Action Items**

- 1. Immediately investigate top 50 high-risk channels
- 2. Review 100 channels with multiple anomalies
- 3. Implement automated filtering for channels with quality score < 3.0
- 4. Set up real-time alerts for new fraud patterns

### **Quality Score Analysis**

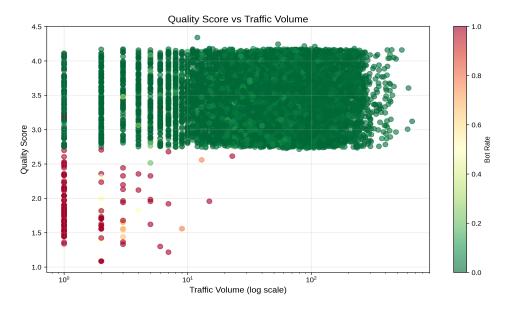
#### **Channel Quality Distribution**



**What this shows:** This histogram displays how channels are distributed across different quality score ranges. The x-axis shows quality scores (0-10), and the y-axis shows the number of channels in each range.

**How to interpret:** Look for concentration patterns: A right-skewed distribution (more channels with high scores) indicates overall good traffic quality. Left-skewed suggests widespread quality issues. Bimodal distributions may indicate distinct channel groups.

#### Quality Score vs Traffic Volume



**What this shows:** This scatter plot correlates channel quality scores with traffic volume. Each point represents a channel, with position indicating its quality (y-axis) and traffic volume (x-axis, log scale).

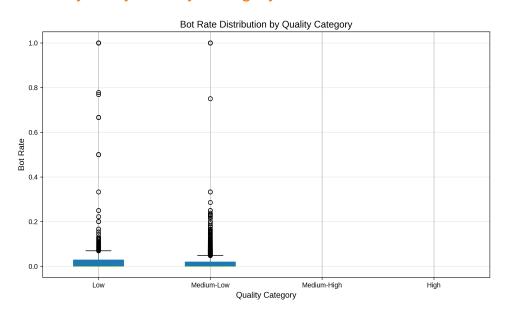
**How to interpret:** Ideal channels appear in the top-right (high quality, high volume). Bottom-right channels (low quality, high volume) pose the highest risk. Top-left channels (high quality, low volume) may be underutilized opportunities.

# **Top Performing Channels**

Channel ID	Quality Score	Bot Rate	Volume
9435	4.34	0.0%	12
9588	4.24	0.0%	54
1391	4.22	1.1%	88
2456	4.21	0.0%	107
5831	4.18	0.0%	15

# Risk Analysis

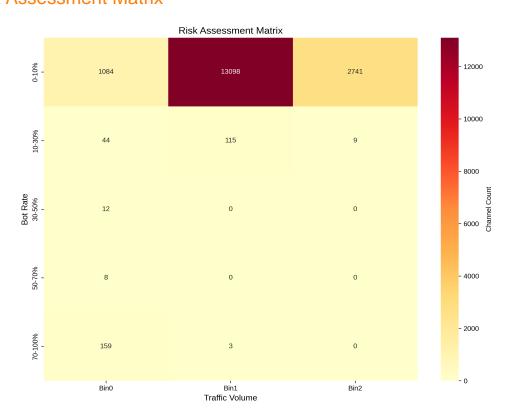
### Bot Rate Analysis by Quality Category



**What this shows:** This box plot shows the distribution of bot rates within each quality category. The boxes show quartiles, whiskers show the range, and dots indicate outliers.

**How to interpret:** Lower bot rates in higher quality categories validate the scoring model. Wide boxes indicate high variability within a category. Many outliers suggest inconsistent patterns that need investigation.

#### Risk Assessment Matrix

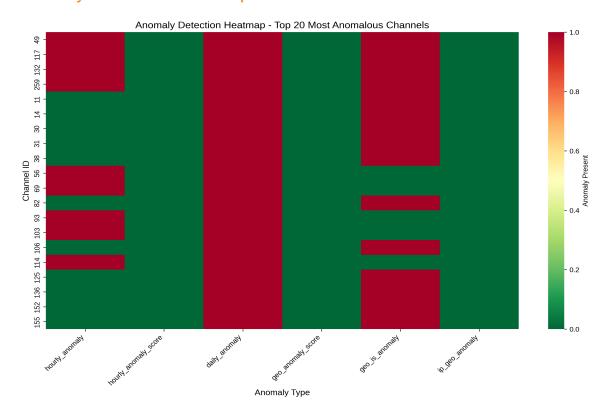


**What this shows:** This 2D heatmap maps channels by bot rate (y-axis) and volume (x-axis), with color intensity showing the number of channels in each zone.

**How to interpret:** Red zones (high bot rate + high volume) require immediate action. The darker the color, the more channels in that risk zone. Focus on reducing channels in the upper-right quadrant.

# **Anomaly Detection Results**

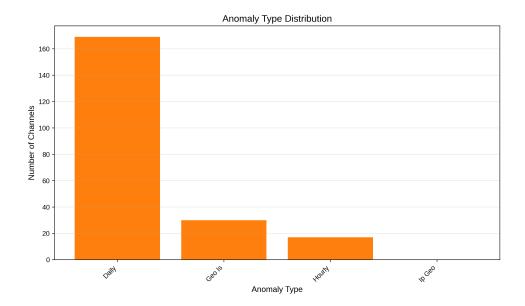
### **Anomaly Detection Heatmap**



**What this shows:** This heatmap visualizes which channels (rows) have which types of anomalies (columns). Dark cells indicate the presence of an anomaly.

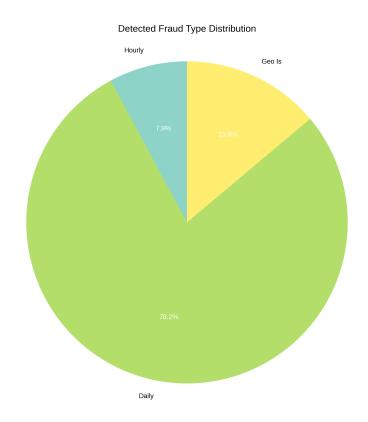
**How to interpret:** Channels with multiple dark cells across the row have multiple anomaly types and need priority investigation. Columns with many dark cells indicate common anomaly patterns affecting many channels.

### **Anomaly Type Distribution**



**What this shows:** This bar chart shows how many channels are affected by each type of anomaly, helping prioritize which anomaly patterns to address first.

**How to interpret:** Focus on anomaly types with the highest counts first, as they affect the most channels. Types with very low counts might indicate rare but potentially serious issues.

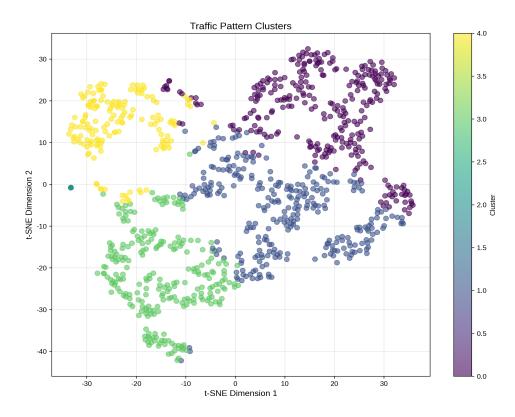


What this shows: This pie chart breaks down the proportion of different fraud types detected across all channels.

**How to interpret:** Large slices indicate prevalent fraud types requiring systematic solutions. Multiple small slices suggest diverse fraud tactics. Use this to prioritize anti-fraud measures.

# Traffic Similarity Analysis

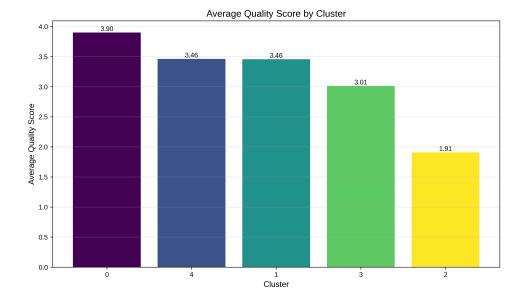
#### **Traffic Pattern Clusters**



**What this shows:** This t-SNE plot shows how channels group into natural clusters based on their traffic patterns. Each point is a channel, colored by its cluster assignment.

**How to interpret:** Well-separated clusters indicate distinct traffic patterns. Channels far from any cluster center are outliers. Large clusters may represent common traffic patterns, while small clusters might be niche or suspicious.

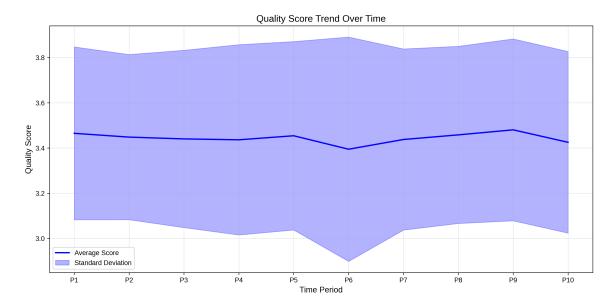
#### Average Quality Score by Cluster



**What this shows:** This bar chart displays the average quality score for each traffic cluster, helping identify which traffic patterns correlate with quality.

**How to interpret:** Clusters with low average quality scores likely contain fraudulent or low-quality traffic patterns. High-scoring clusters represent desirable traffic patterns to encourage.

# **Quality Score Analysis**



**What this shows:** This line chart tracks quality score trends over time periods, revealing patterns and changes in traffic quality.

**How to interpret:** Upward trends indicate improving quality. Sudden drops may signal new fraud campaigns. Seasonal patterns help predict future quality fluctuations.

#### Recommendations

#### **Immediate Actions Required**

- 1. **Block/Investigate High-Risk Channels:** 2624 channels identified as high-risk with average bot rate of 7.9%
- 2. Review Anomalous Patterns: Focus on channels with multiple anomaly types
- 3. Protect Revenue: Potential revenue at risk: \$16951.90

#### **Short-term Improvements**

- 1. Quality Improvement: Work with 14661 medium-low quality channels
- 2. Pattern Monitoring: Set up alerts for channels matching high-risk patterns
- 3. Enforce Standards: Require minimum quality score of 4.0 for new channels

#### **Long-term Strategy**

- 1. Model Enhancement: Retrain models monthly with new data
- 2. Process Optimization: Automate channel blocking for scores < 2.0
- 3. Dashboard: Create real-time dashboard for ongoing monitoring
- 4. ML Integration: Implement real-time scoring for new channels