

2.3. Fraud Detection Techniques #Detection

2.3.1 Intro

Fraud-detection approaches have evolved and gained significant power over the past years.

- Adopting powerful *statistically-based methodologies*.
- Analyzing *massive amounts of data*.

Frauds remain a dynamic phenomenon hard to detect

1. **Fraudsters adapt** their approaches to *commit fraud without being exposed*.

Probe fraud-detection and prevention systems:

- To *understand* their *functioning*.
- To *discover* their *weaknesses*.

2. **Fraudsters develop** advanced strategies to **cover/blend in** their tracks to **avoid being detected** ~ *camouflage*

2.3.2 Fraud-Detection Techniques #Techniques

Need for *new techniques* that are able to **detect and address stealthy patterns**.

1. Unsupervised learning or *descriptive analytics* techniques.
2. Supervised learning or *predictive analytics* techniques.

2.3.2.1 Unsupervised learning techniques or descriptive analytics

Unsupervised: #Unsupervised

- They do **not** require *labeled observations*.
- Learn from historical observation:
Behavior that deviates from normal one = **Detecting anomalies**.

Allow **detecting novel fraud pattern**, not discovered by expert systems since they:

- Are *different* in nature *from historical fraud*.
- Make *use* of *new, unknown mechanisms*.

In the end:

| **Complementary tool** to improve its expert rule-based fraud-detection system.

Limitations: #Limitations

Detect if a *new fraud mechanism* leads to *detectable deviations from normality*.

Prone to deception: *camouflage-like* fraud strategies.

| *Need to be improved by complementing other tools*.

2.3.2.2 Supervised learning techniques or predictive analytics

Supervised: #Supervised

#DEF Learn from historical observations to retrieve patterns that allow differentiating normal and fraudulent behavior.

Aim at *finding "known alarms"*: tracks that fraudsters *cannot* hide.

Can be applied to:

- Predict fraud.
- Detect fraud.
- Estimate the amount of fraud.

Limitations: #Limitations

1. They *need historical examples to learn* from (i.e., a labeled data set of historically observed fraud behavior).
2. *Low detection power against different and new fraud types* (i.e., not detected so far and not included in the historical database of fraud) -> detected by descriptive analytics.

Complementarity of supervised and unsupervised methods:

Use of *both methods* in developing a *powerful fraud-detection and prevention system*, they focus on *different aspects of fraud*.

2.3.2.3 Social network analysis

Extends the abilities of the fraud-detection system by *learning and detecting characteristics* of fraudulent behavior *in a network of linked entities*.

Including an extra source of information in the analysis, being the relationships between entities, it *contributes in uncovering particular patterns indicating fraud*.

2.3.3 Developing a fraud-detection system

1. Expert-based rule engine.
2. Unsupervised learning systems.
3. Supervised learning systems.

The exact order of adopting the different techniques depend on the characteristics of the type of fraud.

Next chapter: [Fraud Management Cycle](#)