

# Workshop 1

## Use Case

# Agenda

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# Agenda

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## Motivation:

Financial Institutions are used by criminals to “legitimize” money generated from criminal activities, hence banks are expected by society and regulators to act proactively in detecting and stopping criminals from using the financial system to access their money through effective monitoring and reporting to authorities.

## Complication:

- **Number of clients** and **transaction volume** are **large** and only expected to grow in the future
- **Diminishing returns of adding more headcount**
- **Criminals** are constantly **innovating** on ways to use **financial system** to “**legitimize**” their funds

## Ask:

Use suitable **data analytics tools and techniques** to help **Scotiabank detect financial crimes**:

- **Find known high risk people in our customer base using public data**
- **Score clients according to their likelihood of being involved in Money Laundering using transactional data**
- **Enhance scoring and visualize networks using connections between clients**

# Agenda

~~01~~ ~~Case Theme~~

**02** **Background**

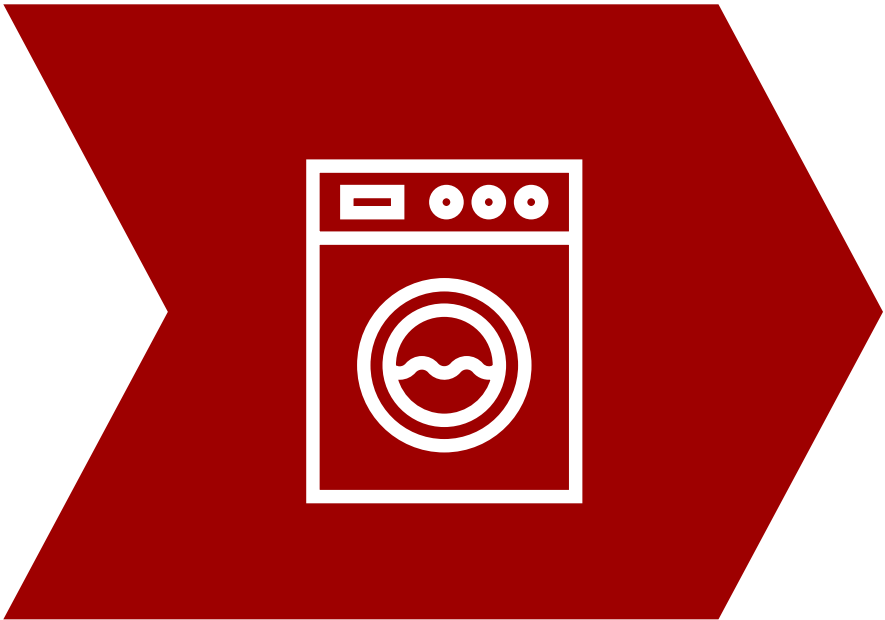
03 Tasks

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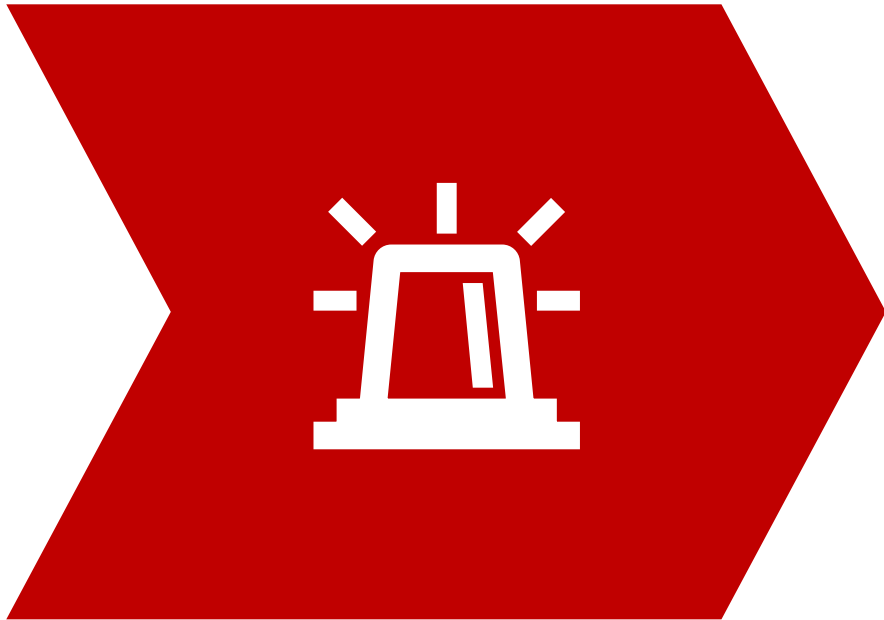


01



What is Money  
Laundering ?

02



How can we prevent  
Financial Crimes?

03



Risk Ratings

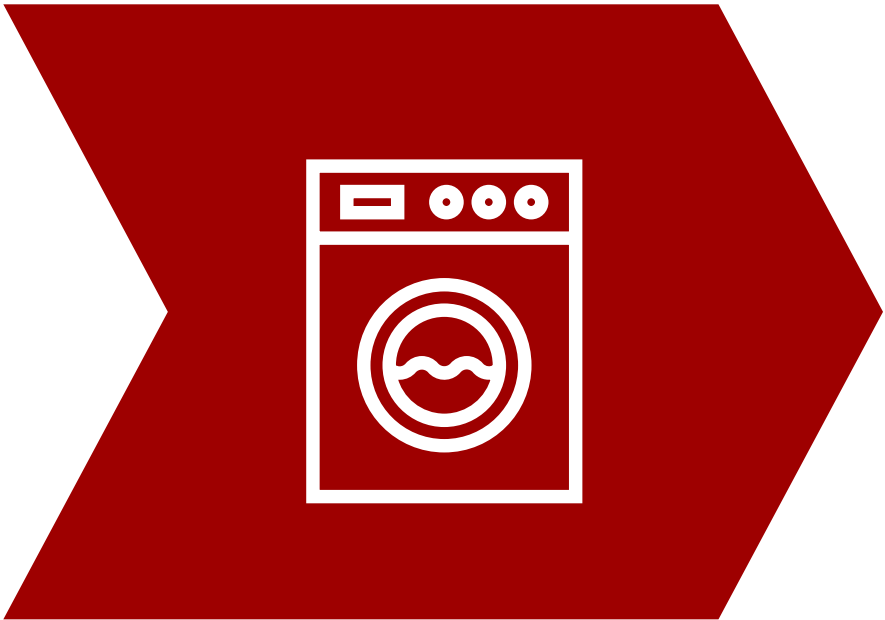
04



Customer  
Connections

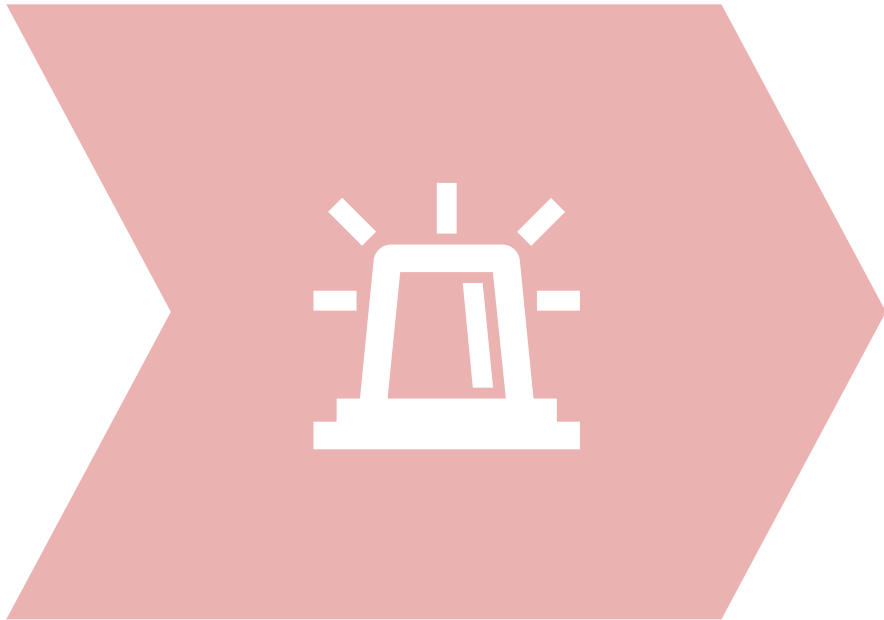


01



What is Money Laundering ?

02



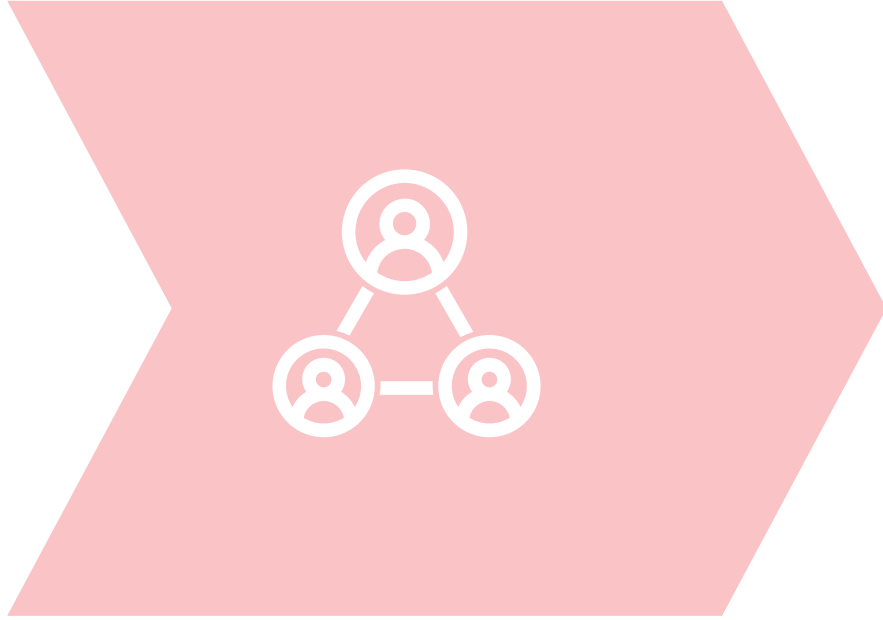
How we can prevent Financial Crimes?

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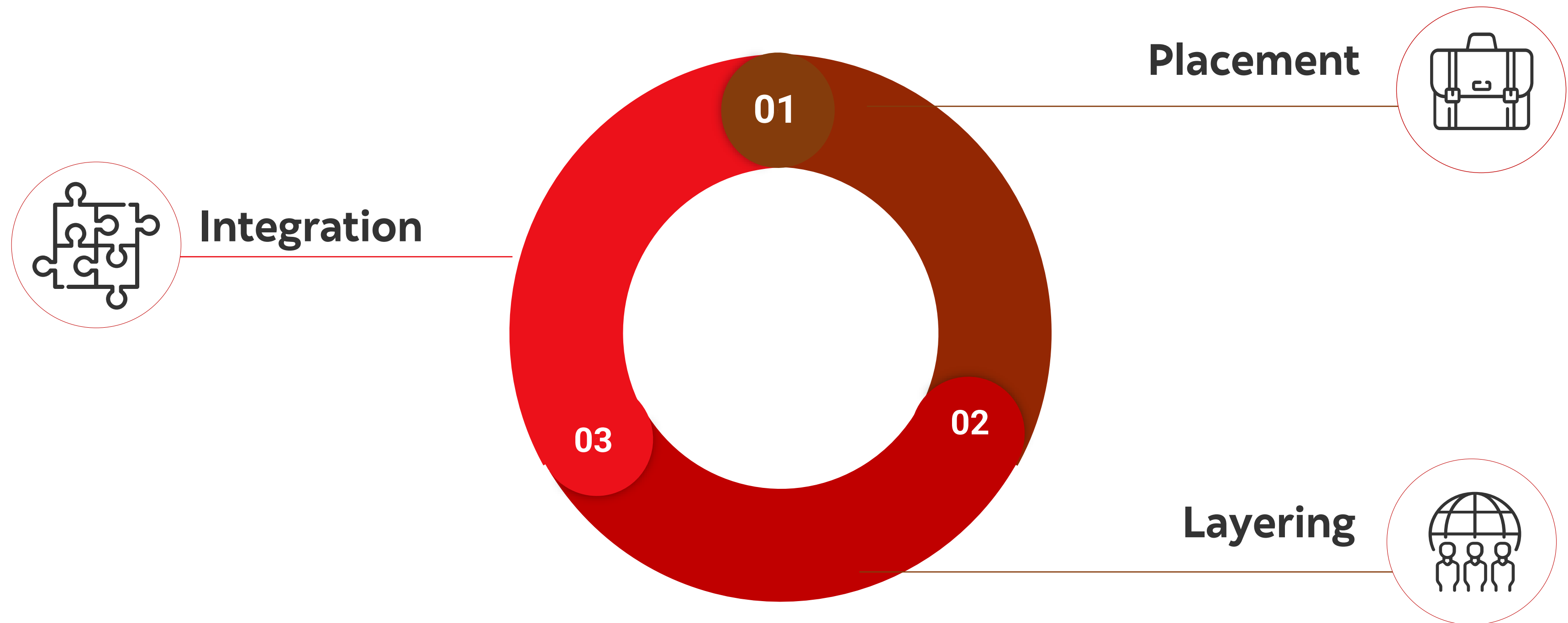
Risk Ratings

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Customer Connections

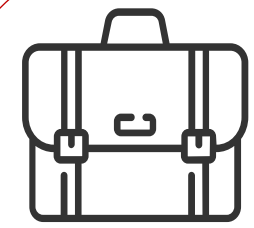
**Definition:** Money laundering is the act of “turning the proceeds of crime into cash or property that looks legitimate and can be used without suspicion.”





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## Placement



01

### Stage 1: Placement

Money laundering starts by placing funds derived from illegal activities into the financial system in the form of **cash deposits, cheque, money transfers or any type of transaction where money enters the Bank.**

**Placement can be identified at the frontline by understanding the source of the funds.**

**Example:** customer deposits a total of **\$20,000 in cash** using **multiple transactions** to **avoid reporting and detection** of large cash deposits



**Definition:** Money laundering is the act of “turning the proceeds of crime into cash or property that looks legitimate and can be used without suspicion.”

## Stage 2: Layering

Second stage is for a criminal to move **funds around to hide their illegal origins**. It may consist of **multiple transactions without clear purpose** to move money between **products, clients, bank, corporations and geographies**

**Layering can be detected by understanding transaction purpose and if it is unusual for the customer.**

### Examples:

- Customer receives an incoming wire from Country A for \$20,000, then sends an outgoing wire in the same amount to Country B, or
- Customer makes a deposit to their account, then transfers the funds to another of their accounts with another financial institution

Placement

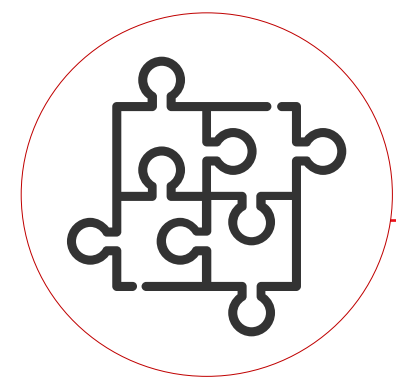


02

Layering



**Definition:** Money laundering is the act of “turning the proceeds of crime into cash or property that looks legitimate and can be used without suspicion.”



## Integration

### Stage 3: Integration

The final stage **integrates the illicit proceeds**, that now appear to be clean funds, **into the economy as "normal"** personal or business transactions. By this stage is **hard to distinguish between legal and illegal money**, and the **criminals** can now **use these funds without suspicion**.

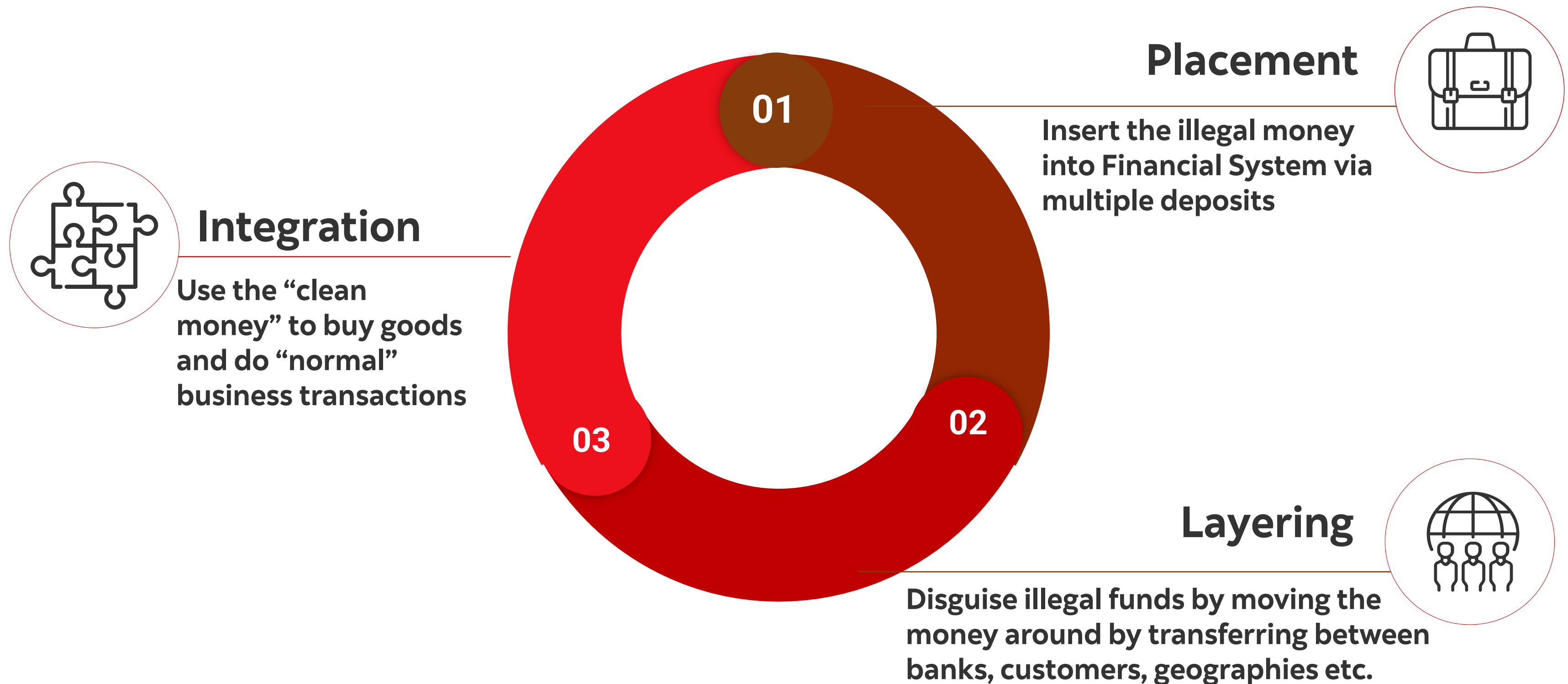
**Integration can be detected by knowing the client and monitor for odd and unusual transactions.**

#### Examples:

- Use illicit proceeds to purchase luxury assets such as real estate, high end artwork, jewelry or vehicles
- A client opens an offshore bank account with a debit or credit card to access the account. They wire illicit proceeds to this offshore account, which are used for withdrawals or purchases

03

**Definition:** Money laundering is the act of “turning the proceeds of crime into cash or property that looks legitimate and can be used without suspicion.”



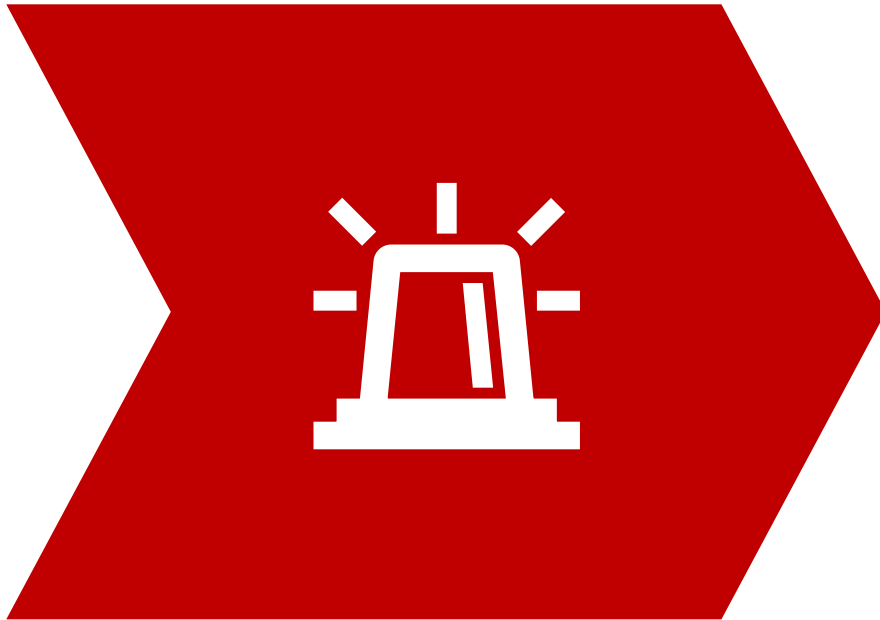


01



What is Money Laundering ?

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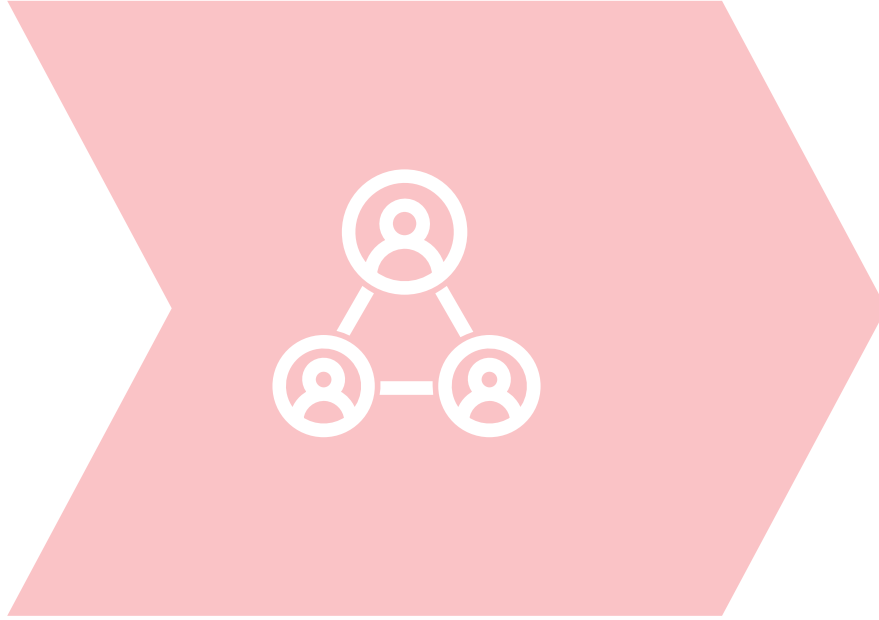
How we can prevent Financial Crimes?

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Risk Ratings

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Customer Connections



**Know Your Customer (KYC)**

**Ongoing Monitoring**



**Know Your Customer (KYC)**

Ongoing Monitoring



## Know Your Customer (KYC)



### Collect Data

**Accurately collect all (KYC) information at on-boarding**

- Name, Address, Occupation
- Nature of Business
- Transactional Counterparties
- Source of funds and wealth
- Account activity
- Product Usage



### Name Screening

**Systematically check new clients name against watchlists**

- Prohibited of conducting business with**
- Known criminals
  - Terrorists
  - Sanctioned parties and countries

**Clients that match any criteria will not be onboard or; assets and accounts will be frozen**



### Client Risk Rating

**Initial Risk Rating is determined at onboard and is either: Low, Medium or High**

It is computed on risk factors and takes into account several customer data points, such as KYC information.

Ratings are not static and can change throughout the lifecycle as more information is gathered about the client.





~~Know Your Customer (KYC)~~

**Ongoing Monitoring**



## Ongoing Monitoring



### Up-to-date Data

**Keeping client information up-to-date**

- Recurrently update customer information
- Reassess client data when changes occur, such as opening new accounts, requesting products, adding account holder, change of address etc.



### Transaction Monitoring

**Systematically monitor transactions volume and frequency**

- Use historical data and analytics to deploy automated models to detect and flag abnormal behaviour
- File for further investigations if an alert is raised
- Cash transactions > \$10k are reported to FINTRAC



### Payment/Name Screening

**Continuously monitor client names and payment data with most updated Watchlist**

- Watchlists and news are constantly developing, and the bank must be vigilant to detect any recent criminal activity going on
- Use NLP techniques to compare names, extract funds source etc.

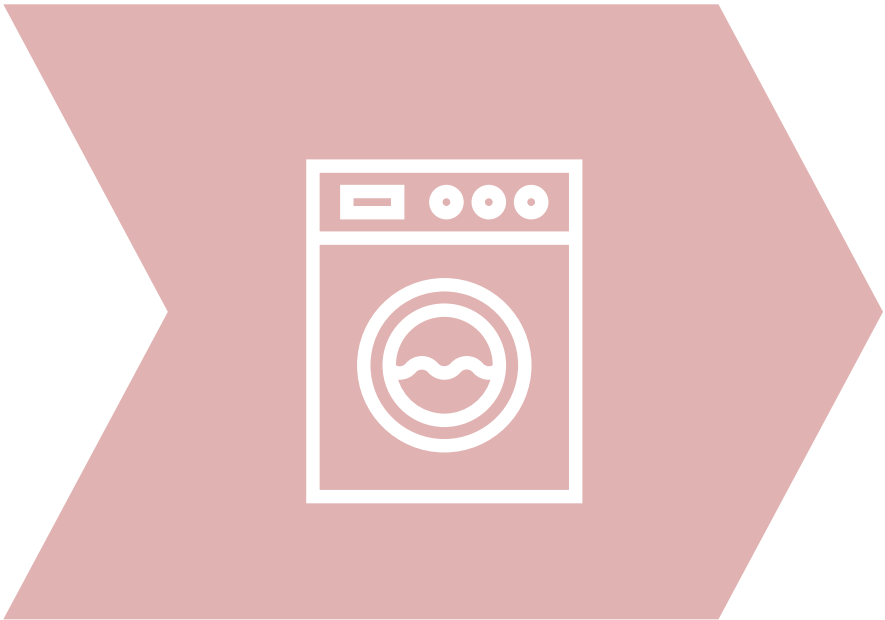


~~Know Your Customer (KYC)~~

~~Ongoing Monitoring~~

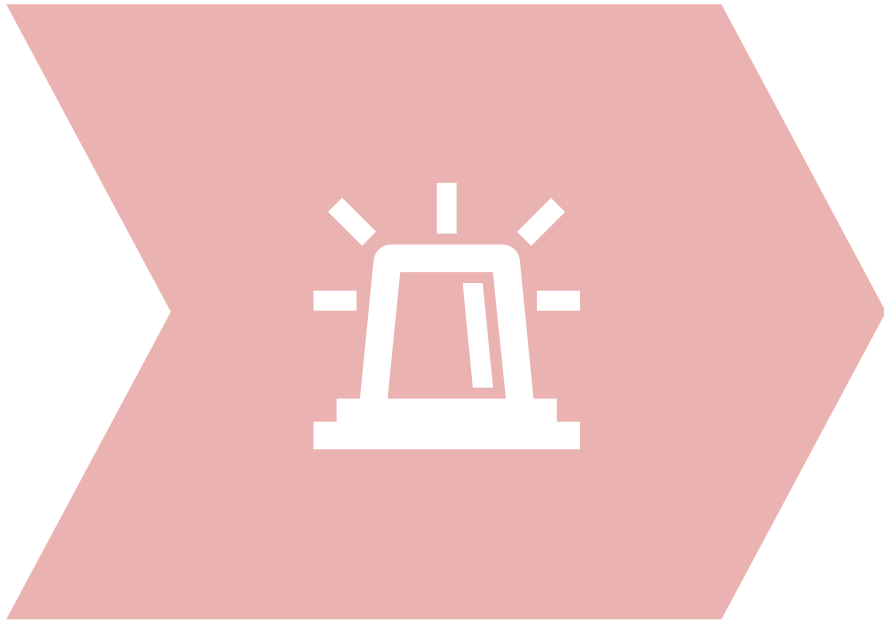


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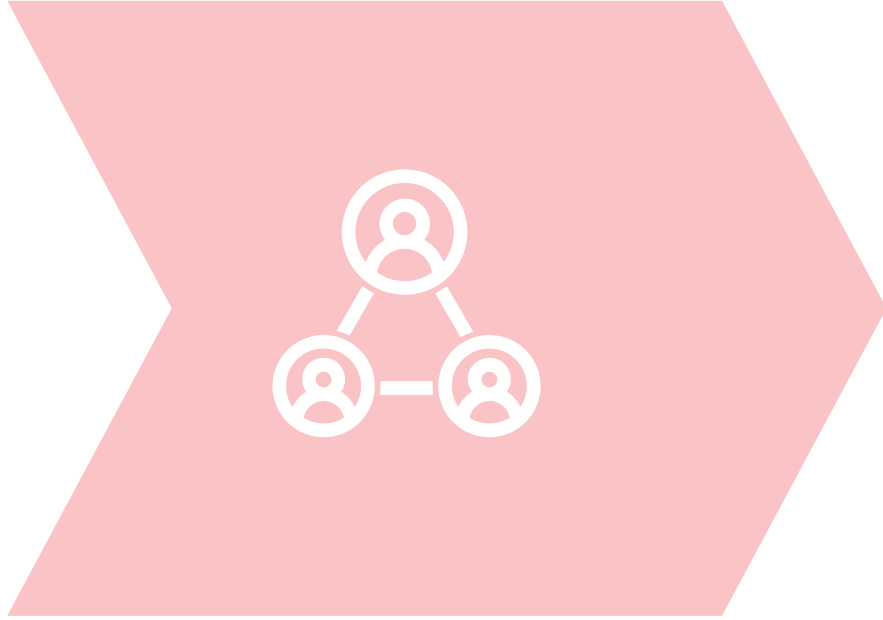
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Risk Ratings

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Customer Connections

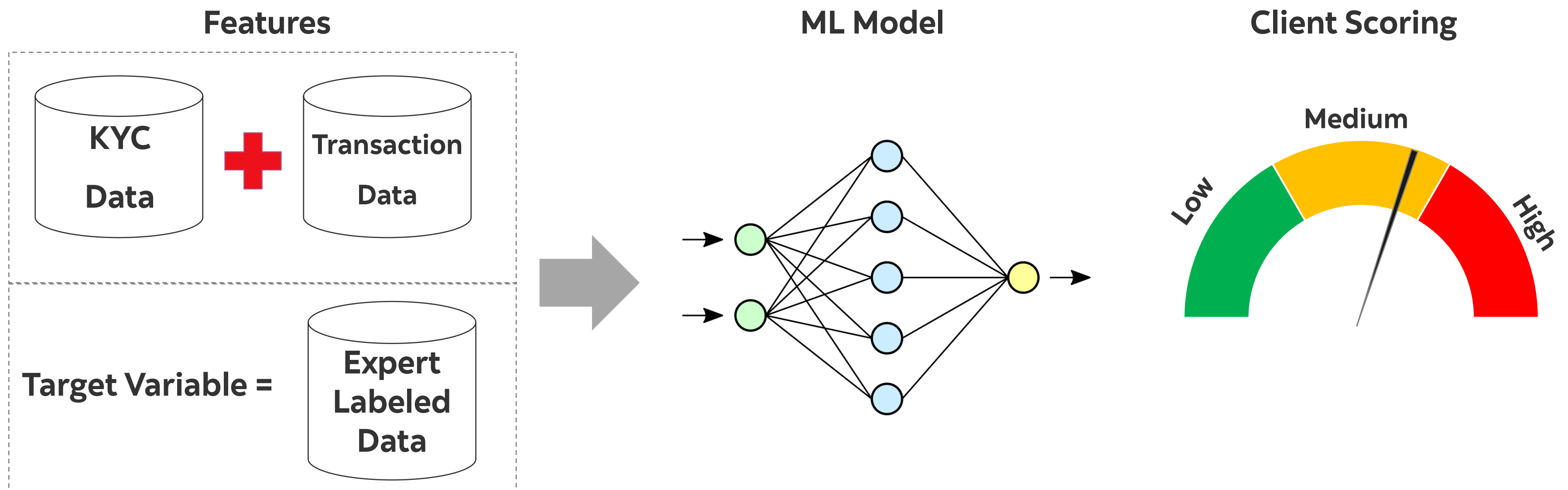
**Definition:** Risk Ratings is a scale that represents how likely a client is involved in money laundering activities.

**High Risk: very high likelihood** of engaging in money laundering

**Low Risk: highly unlikely** of being involved in financial crimes

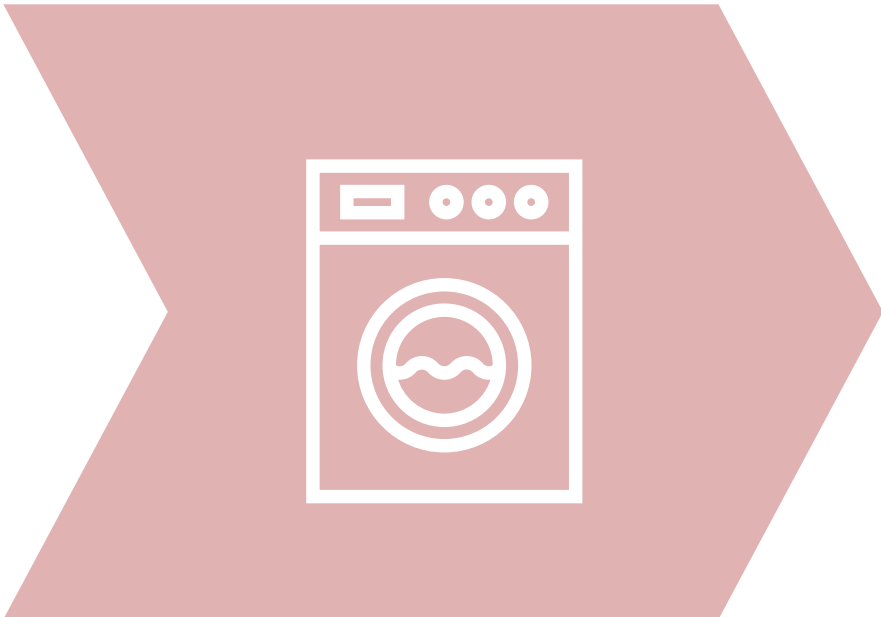
**Medium risk:** anywhere between low and high risk

**How ratings are computed:**



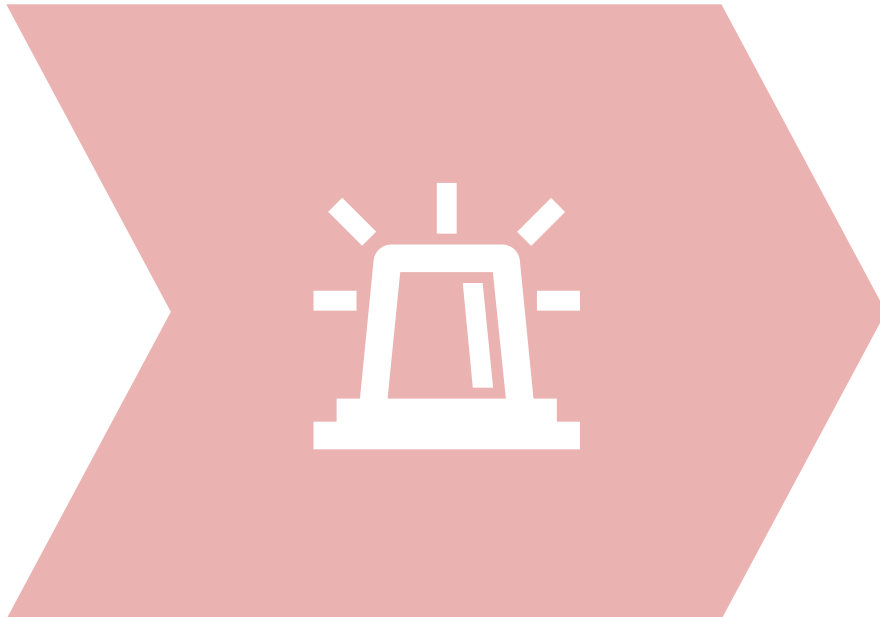


01



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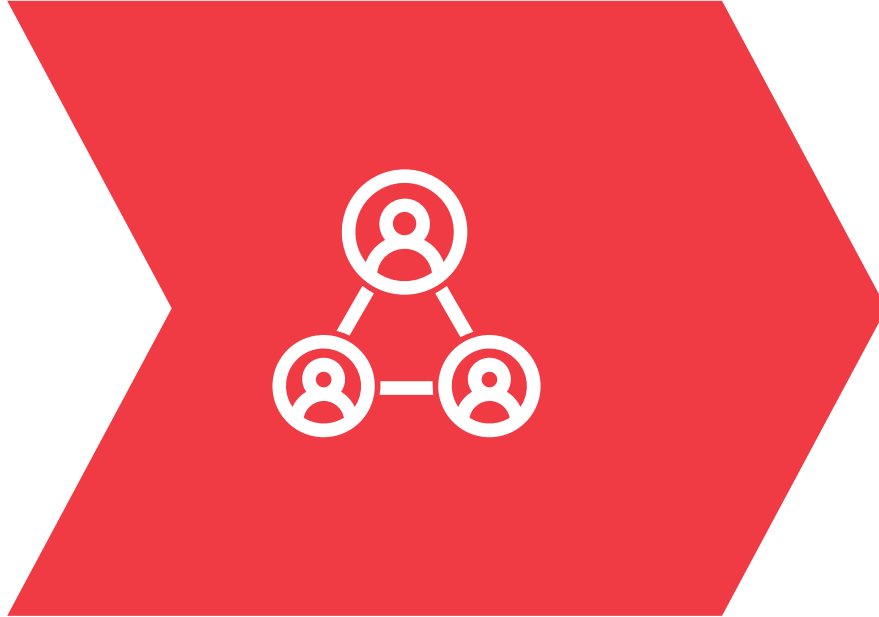
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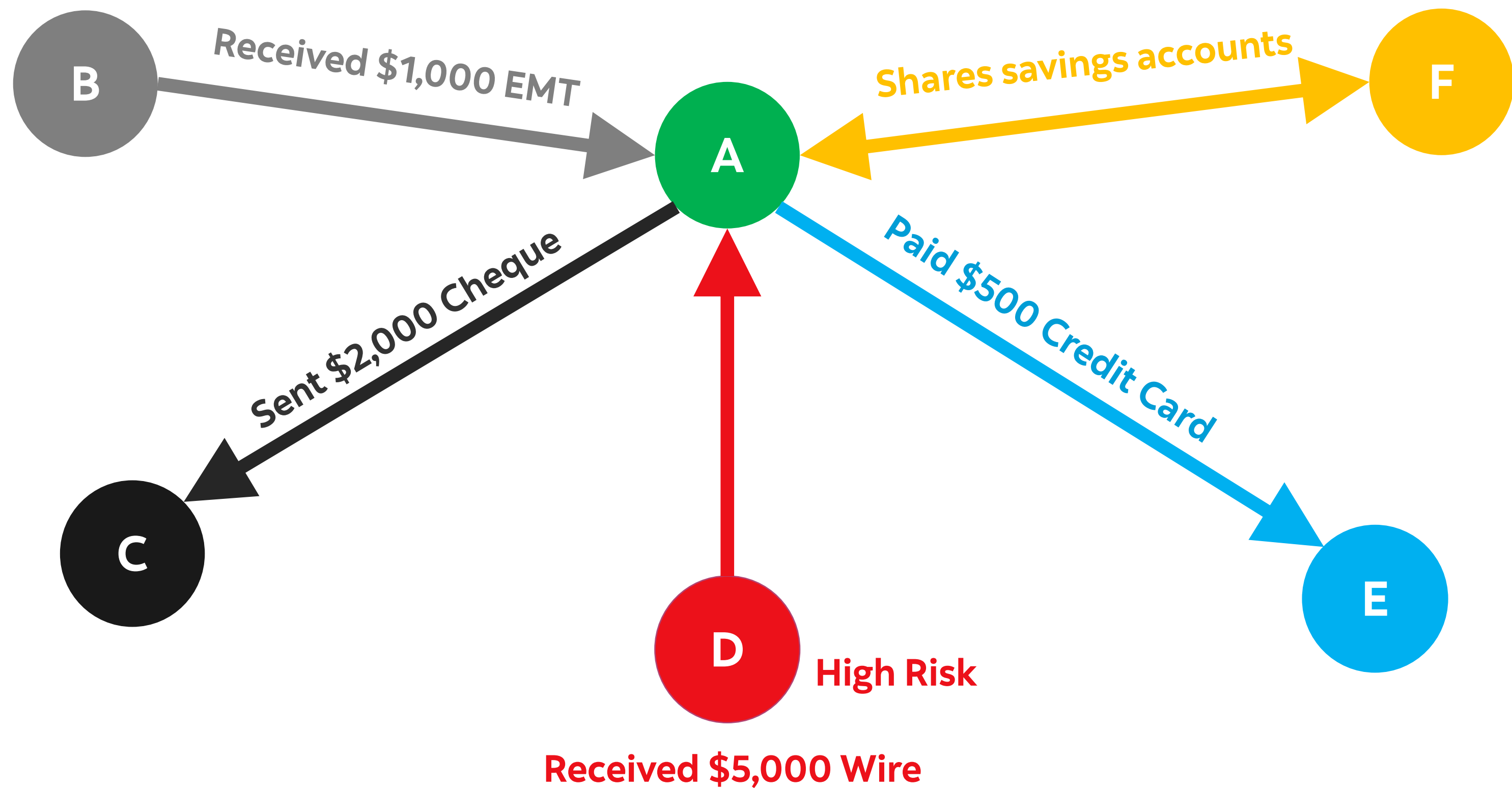
Risk Ratings

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Customer Connections

**Definition:** customer connection is any relationship between clients, either shared account ownership, customer information (same address, email, phone number etc.) or **money transfer**



# Agenda

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**03** **Tasks**

04 Data Sources

05 Summary



**1**

## **Name Screening**

**Detect 50 Bad actors in our customer base using public data sources**

**Ask:**

**Find as many bad actors as possible using NLP techniques to match customer names with watchlist and other relevant information given**

**2**

## **Risk Rating**

- A. Classify customers into Low, Medium and High risk.**
- B. Predict Bad actors using results from Task 1 as target variable**

**Ask:**

**Using transactional and KYC data, create models to**

- A. Classify clients into Low, Medium or Risk**
- B. Predict bad actors using your results from Task 1 as dependent variable**

**3**

## **Improve model using Graph data**

**Add customer connections information to improve Task 2 models or use a graph model directly**

**Ask:**

- Improve your Task 2 models by using client connections to either:**
- Extract new features**
  - Fit graph models directly**
  - Visualize interesting networks**

**Task 1: Name Screening**

**Task 2: Supervised Learning**

**2A: Risk Rating**

**2B: Predicting Bad Actors**

**Task 3: Improve your models**



## **Task 1: Name Screening**

### Task 2: Supervised Learning

#### 2A: Risk Rating

#### 2B: Predicting Bad Actors

### Task 3: Improve your models

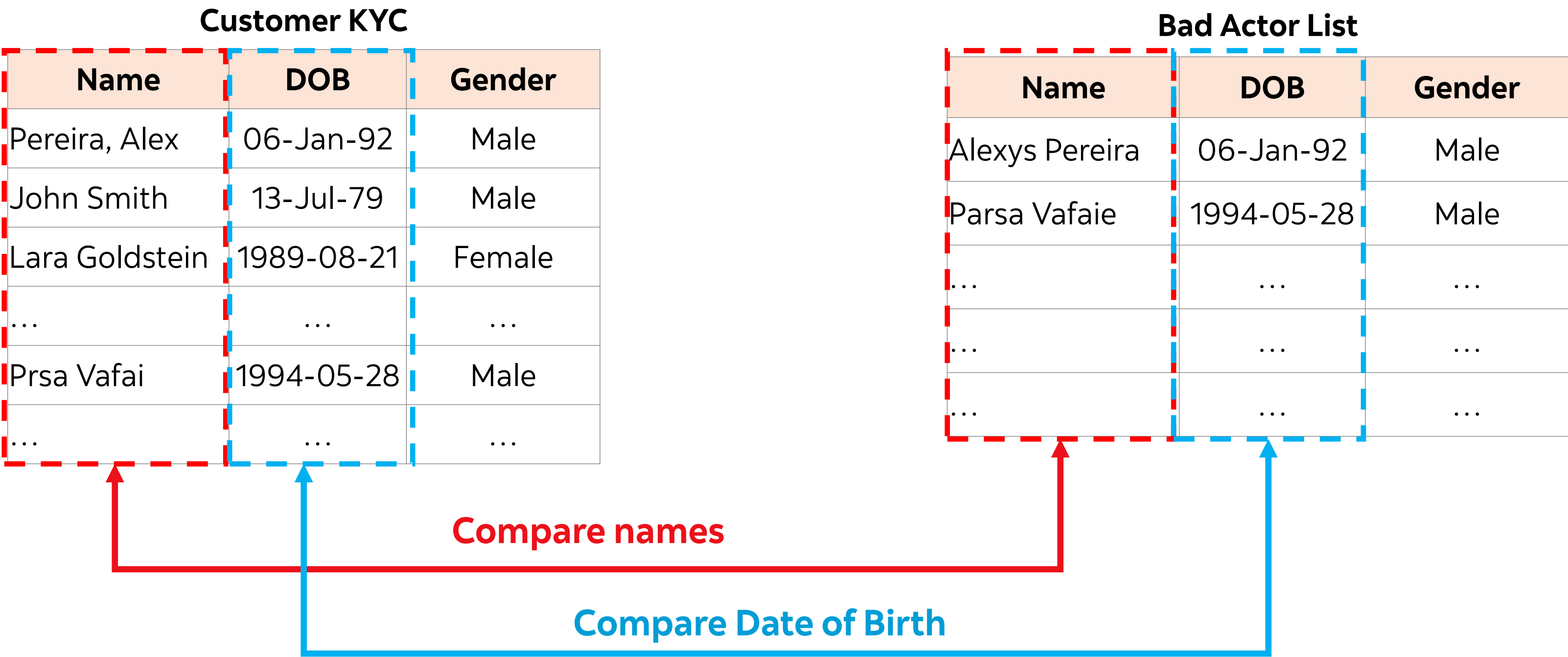


**Problem:** there is a total of 50 bad actors in our customer base that were extracted from OpenSanctions (watchlist), amongst them we have warmongers, sex trafficker, drug dealers etc.

**Ask:** using appropriate NLP techniques for name matching and other relevant KYC data, find as many bad actors as possible.

*Hint: formulate a similarity scoring methodology, and assign bad actors based on a cutoff*

**Example:**





~~Task 1: Name Screening~~

## **Task 2: Supervised Learning**

2A: Risk Rating

2B: Predicting Bad Actors

~~Task 3: Improve your models~~



~~Task 1: Name Screening~~

## **Task 2: Supervised Learning**

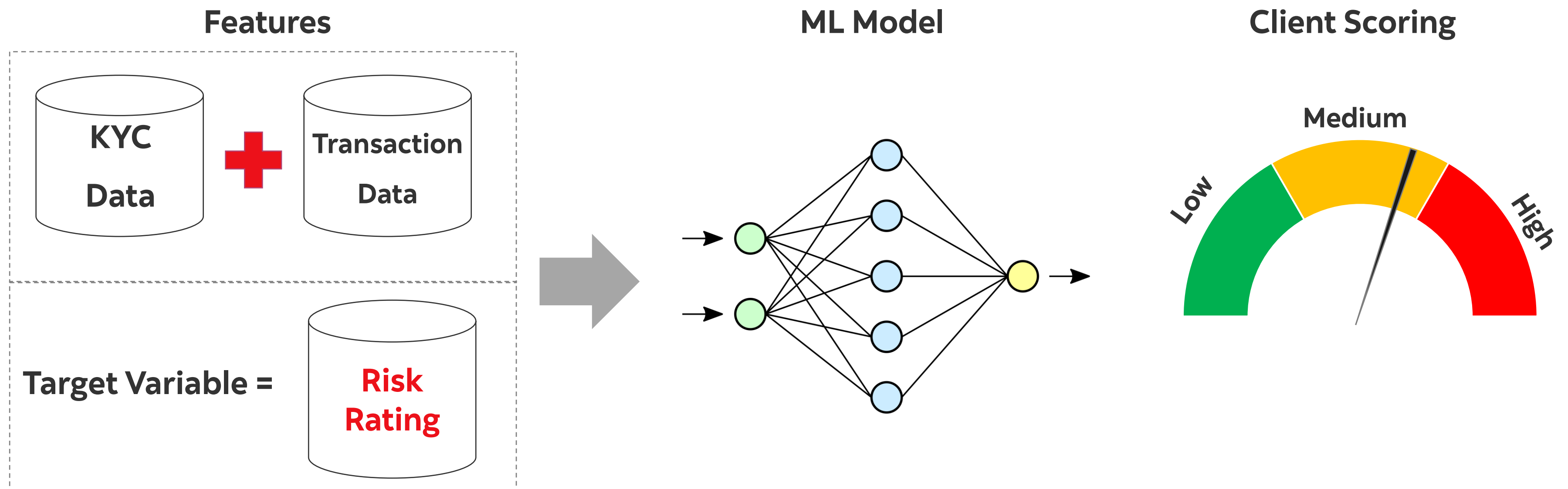
### **2A: Risk Rating**

2B: Predicting Bad Actors

Task 3: Improve your models

**Problem:** there are too many clients in our customer database and heterogenous, we would like to automate this process to be as efficient and reliable as possible

**Ask:** using predictive modelling classify all 1M customers into three risk buckets: Low, Medium and High. As features you must use KYC data and transactional data, your target variable is the risk rating provided



**Similar to the Risk Ratings presented earlier, we would like you to produce your own solution to this problem**



~~Task 1: Name Screening~~

## **Task 2: Supervised Learning**

~~2A: Risk Rating~~

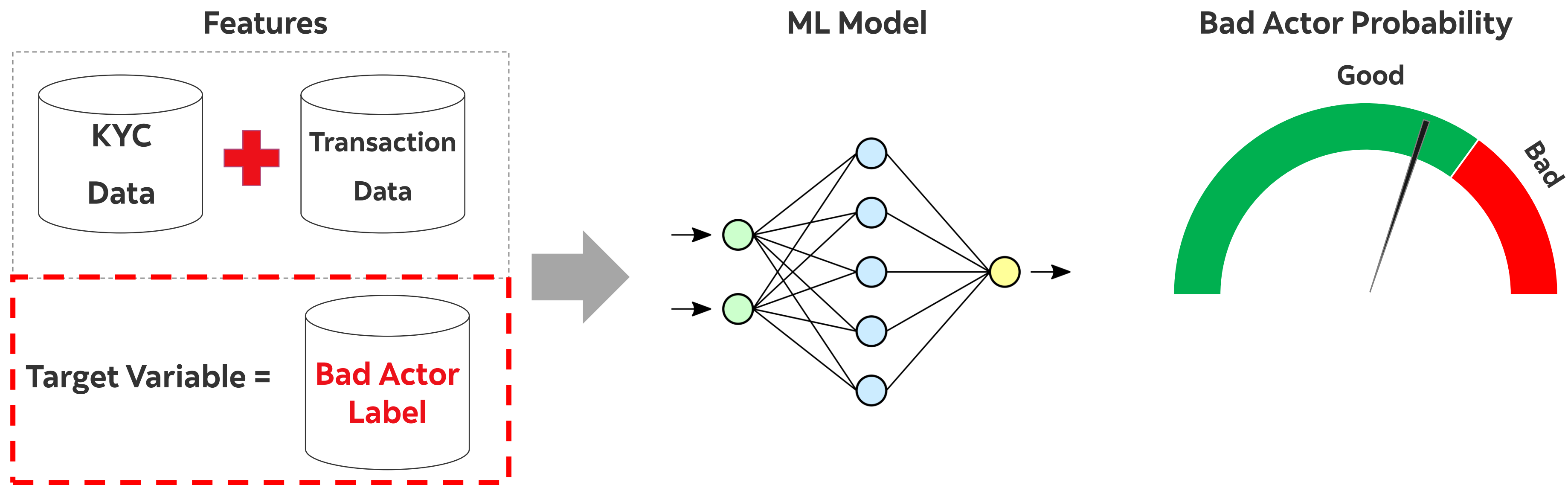
### **2B: Predicting Bad Actors**

~~Task 3: Improve your models~~



**Problem:** bad actors are hard to detect, and their presence is a high risk to the bank, we also would like to have an automated process for screening

**Ask:** similar to Task 2A you must use KYC and Transactional data to produce a model to estimate the likelihood of a customer being a bad actor based on your results of Task 1.



**The difference between 2A and 2B is the target variable: replace Risk Rating with your own Bad Actor Label**



~~Task 1: Name Screening~~

~~Task 2: Supervised Learning~~

~~2A: Risk Rating~~

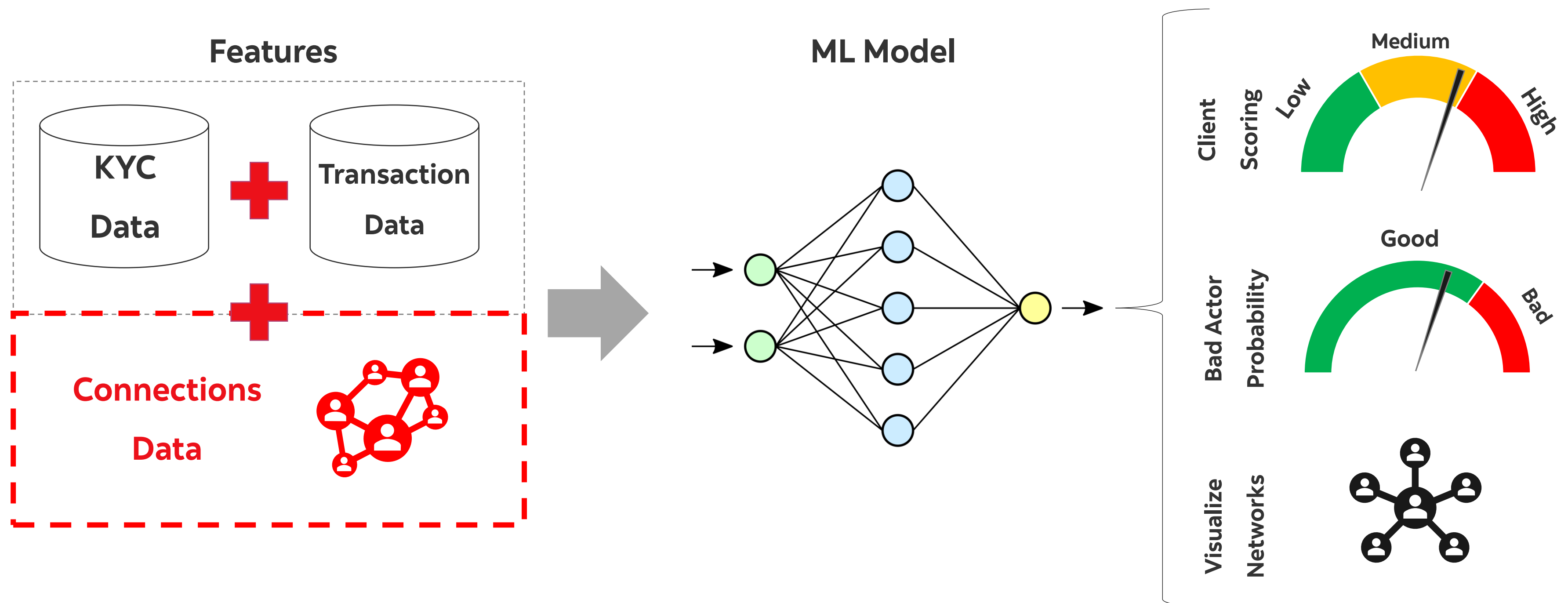
~~2B: Predicting Bad Actors~~

**Task 3: Improve your models**

**Problem:** KYC and Transactional data provide useful information but fails to capture one of the most important components of detecting financial crimes – **money flow**

**Ask:** improve your Task 2 models by using client connections data to either

- a. Engineer new features
- b. Build a Graph model i.e. model that ingests graph data directly
- c. Visualize and detect interesting networks using unsupervised learning





~~Task 1: Name Screening~~

~~Task 2: Supervised Learning~~

~~2A: Risk Rating~~

~~2B: Predicting Bad Actors~~

~~Task 3: Improve your models~~

# Agenda

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**04** **Data Sources**

05 Summary

**Overview:** you will be given 3 Scotiabank data sets and one open-source data for Bad Actors matching

**Open Source:** you should explore the following link to get a list of Bad Actors  
<https://www.opensanctions.org/datasets/default/>

**Scotiabank Data:** synthetic data sets generated using real bank data for this case competition

- **Nodes Data:** main data source for this competition, contains KYC, Transactional data and Risk Rating
  - File name:** UofT\_nodes.csv
  - Number of Rows:** 1M, one row per CUSTOMER\_ID
  - Number of Columns:** 20 columns (1 target column); RISK = Target variable for Task 2
- **Edges Data:** shows the connections between clients i.e. amount of money sent via EMT from one customer (source) to another (target)
  - File name:** UofT\_edges.csv
  - Number of Rows:** 466k
- **Occupation Data:** lookup table that maps an occupation (code) to their risk level of being involved in financial crimes
  - File name:** UofT\_occupation\_risk.csv
  - Number of rows:** 841

# Agenda

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**05** **Summary**

**Overview:** use appropriate modelling techniques to name screen bad actors and classify customers according to their likelihood of being involved in financial crimes using KYC, Transactional and client connections data

**Data:** Three synthetic Scotiabank data sets + One open-source data for Bad Actors matching

- **Open Source:** <https://www.opensanctions.org/datasets/default/>
- **Nodes Data:** KYC + transactional data - UofT\_nodes.csv
- **Edges Data:** client connections - UofT\_edges.csv
- **Occupation Data:** lookup table - UofT\_occupation\_risk.csv

## Tasks:

- **Task 1 (Name screening):** find bad actors in Nodes data using OpenSanctions data
- **Task 2 (Supervised Learning):** use Nodes Data + Occupations data to build models
  - **2A (Risk Ratings):** assign each client to a risk bucket Low, Medium or High
  - **2B (Bad Actors):** compute the likelihood a client being a Bad Actor - **use your Task 1 output as target variable**
- **Task 3 (Improve model):** add Connections Data to Task 2 models or build a model that ingests Graph Data natively.



- 1. Get access and familiarize yourself with the data provided**
- 2. Do an extensive Exploratory Data Analysis to understand correlations and patterns**
- 3. Research papers that use machine learning/statistical techniques to solve similar problem – leverage UofT library access, seek for insights from other domains other than Finance**
- 4. Define a robust validation methodology to assess your models' performance**
- 5. Select relevant evaluation metrics for each problem**
- 6. Start with simple models, learn from their outputs and leverage those insights to:**
  1. Understand the patterns in the data
  2. Draw sensible conclusions for your presentation
  3. And/or have ideas for more complex models
- 7. Plan yourself ahead of time:**
  1. Define by when you want to have presentation and video ready for submission
  2. Make weekly goals to keep momentum going
  3. Split tasks among team members – everyone has something to contribute, play to their strengths

**Disclaimer: we are interested in sensible, robust methodologies, insights derived and how well you communicate your work to an audience.**

# Agenda

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**Q & A**

**Thank you**  
**Good Luck!**