



# DATA AND ANALYTICS FOR ENHANCED CUSTOMER KNOWLEDGE AND GREATER VALUE CREATION

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# Agenda

**CONTEXT & INTRODUCTION**

**ARTIFICIAL INTELLIGENCE TO ENHANCE RISK MANAGEMENT**

**EVIDENCE FROM REAL APPLICATIONS**

**WRAP UP**

# AN EVOLVING MARKET: THREATS AND OPPORTUNITIES

## The need of a new risk management paradigm

- > The market has been characterized for some time by a sort of *general instability* or a *continuous evolution*: competition, regulation, customer behaviours, technology, etc. are all **changing or affecting the rules of the game**
- > In order to «play» the game, institutions need to equip themselves with more adequate solutions: i. gain **greater customer insight** and ii. make more **effective decisions** (e.g. what to take on, what terms, how, etc.)
- > **Technology** is one of the key cards to play as it enables a more and more holistic view of the customers. And it is **moving fast!**

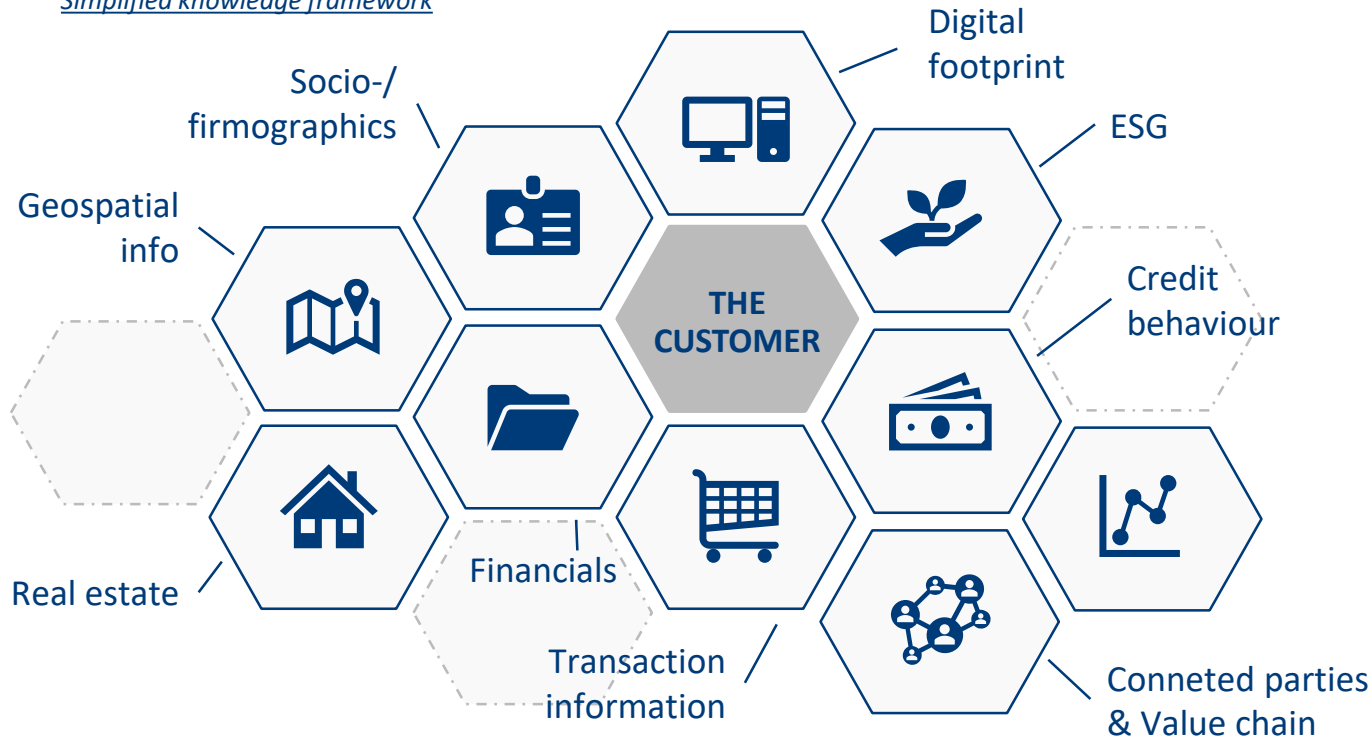
Market forces and trends, illustrative



# HOW MUCH DO AND CAN WE KNOW ABOUT A CUSTOMER?

Identify and analyse the customers' multiple facets: a simplified framework

*Simplified knowledge framework*



## A new paradigm

- > **New structured and unstructured data** is now also available (in addition to traditional data)
- > **Artificial intelligence (from ML to Gen-AI)** is the game changer: data extraction and effective data management

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# ARTIFICIAL INTELLIGENCE: THE ENABLER

Extract and model information



## AI Artificial Intelligence

A discipline that studies how to **simulate**, through a computer, **the decision-making and learning system of human intelligence**



## ML Machine Learning

**Algorithms and statistical models** used by computer systems aimed at machine **learning through** the use of **large amounts of data**



## DL Deep Learning

Neural network **algorithms** that simulate the behaviour of the brain to **solve complex tasks** by learning from huge amounts of data

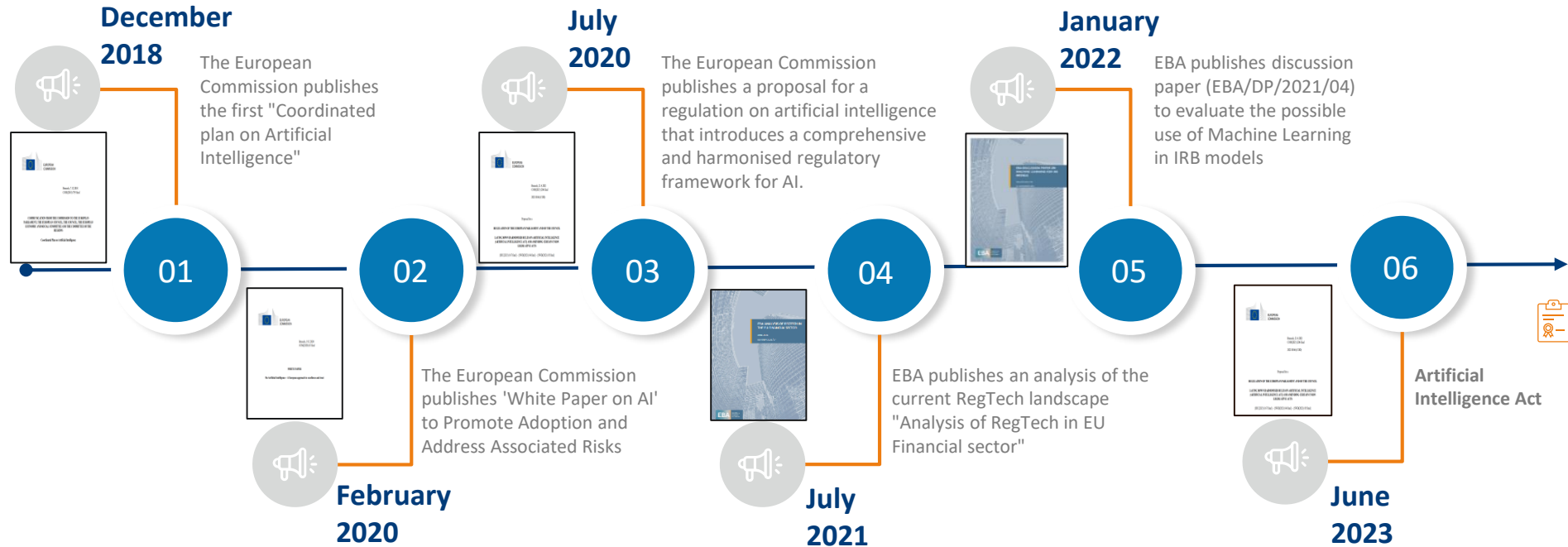


## Gen-AI Generative Artificial Intelligence

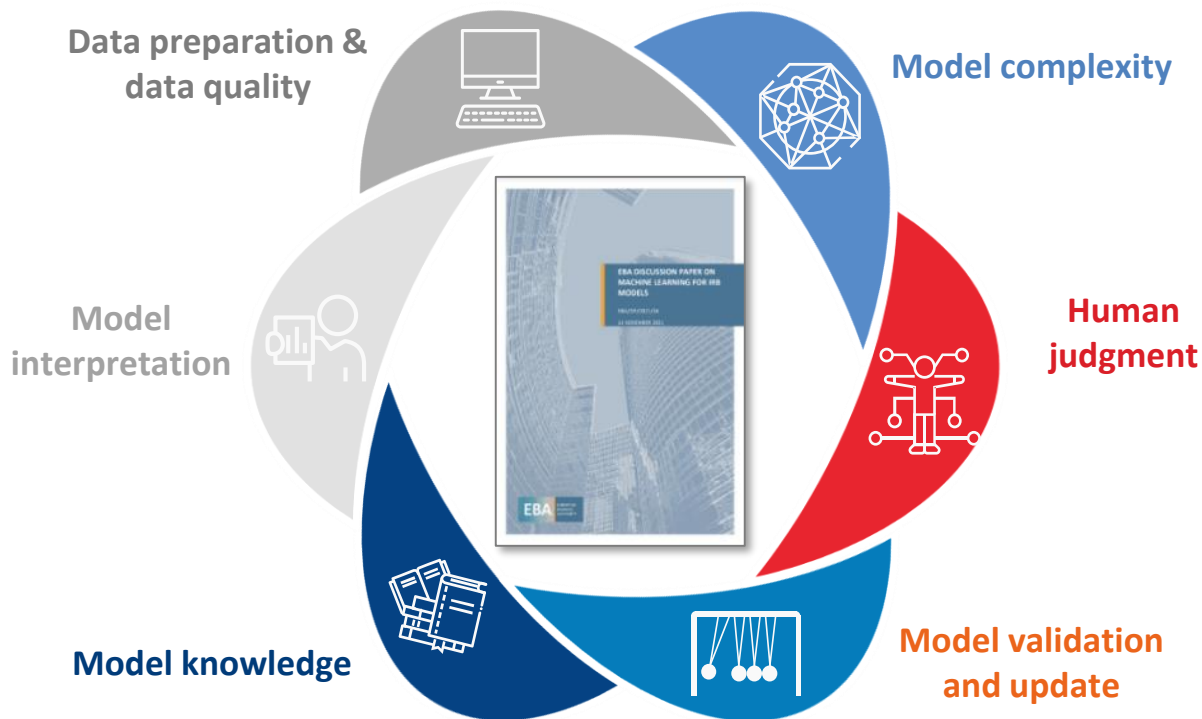
DL models capable of **generating new observations** that have the same properties as the real data used to generate them

# ARTIFICIAL INTELLIGENCE REGULATION IN BANKING

European supervisory authorities' path to assess the impact of AI in the banking world



# USING ML IN RISK MANAGEMENT: SOME CONCERNS



## *Debunking the myth*

ML poses both methodological and regulatory challenges, mostly when applied to credit risk.

Both industry players and supervisors have become more and more familiar with such methodologies, both accepting them (when properly used)



# INDUSTRY APPLICATIONS OF ML IN RISK MANAGEMENT

Discussed with  
the EBA

**MACHINE LEARNING FOR CREDIT RISK  
MANAGEMENT AND IRB MODELS:  
LESSONS FROM SUCCESSFUL CASE HISTORIES**

*A joint paper by Intesa Sanpaolo and CRIF*



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Case study	Benefits
<b>A new IRB model aimed at retail SMEs</b>	Provides customers with a full digital experience, focusing on high digitalisation standards and profiting from PSD2 and the GDPR. Deals with unprecedented conditions, e.g. due to Covid-19
<b>A challenger model to validate the IRB model for retail PDs</b>	Performs initial validation to get supervisory clearance, and ongoing validation to promptly highlight any issues emerging from the bank's model
<b>An early warning system based on transaction data</b>	Ability to use transaction data to identify new information patterns
<b>Second-level controls on a retail SME portfolio</b>	Provides credit analysts, performing credit file reviews, with better (risk-based) sampling criteria and additional information. Allows portfolio-wide analyses
<b>Optimising credit policies</b>	Further automates the "credit policies" used to translate credit scores into credit decisions. Cuts costs and increases speed.

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**CASE 1 – EARLY WARNING SYSTEM WITH TRANSACTION DATA**

WRAP UP

# EARLY WARNING SYSTEM WITH TRANSACTION DATA

## Tier 1 banking group



### BACKGROUND

- > Following JST's finding on the existing early warning system, the Bank decided to evolve the solution embracing all new opportunities offered by technology and methodology: in addition to traditional data assets, transaction data is also made available in a *usable* format (i.e. transactions are categorized)

### SOLUTION

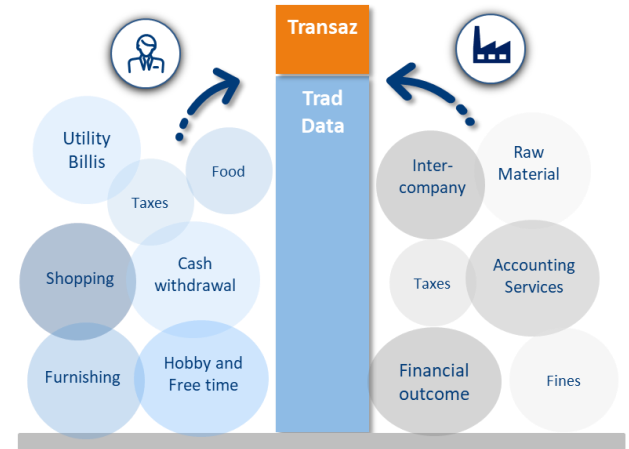


- > **AI-based categorization engine** to label each transaction, grouped by type (macro and micro categories): **millions of transactions/month, 20k+ indicators per customer**
- > **ML-based model development**: from data preparation to model estimation (3 ML model pipelines developed for each segment)



### GOAL

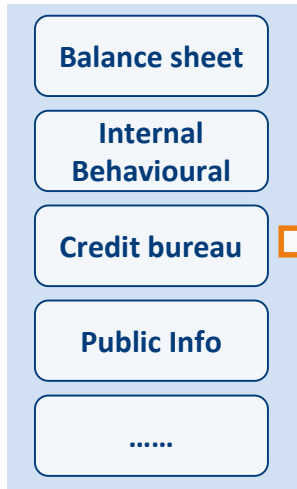
- > **Enhanced model performance** (dealing with JST findings) by **exploiting value of additional available data assets**
- > **Gain confidence about the (ML) solution** ensuring full understanding of development process and model interpretability



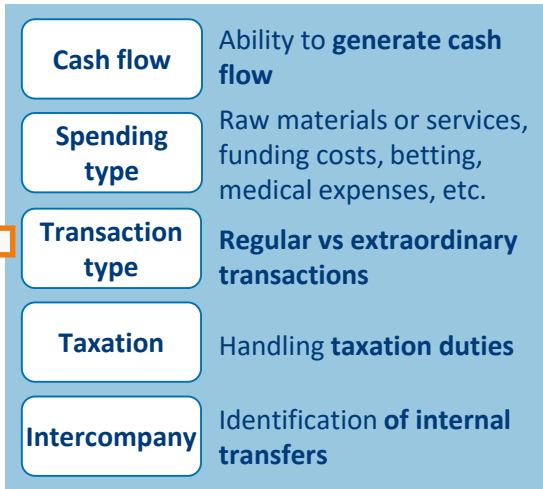
# EARLY WARNING SYSTEM WITH TRANSACTION DATA

## Categorization and modeling of transaction data

### Traditional Information



### Transaction info

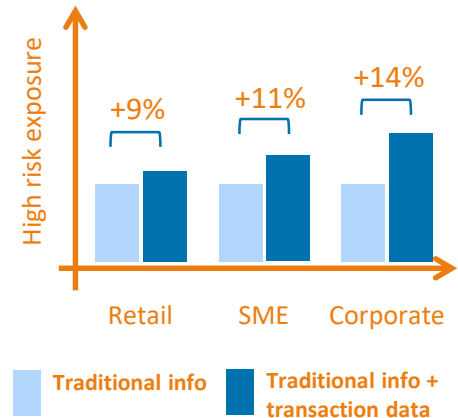


### Signals

- Stability, reactivity to the economic cycle...
- Anomalies or unexpected trends in purchases
- Capturing irregular expenses or revenues
- Management of tax payments
- Netting intra-company transactions

## Real case benefits

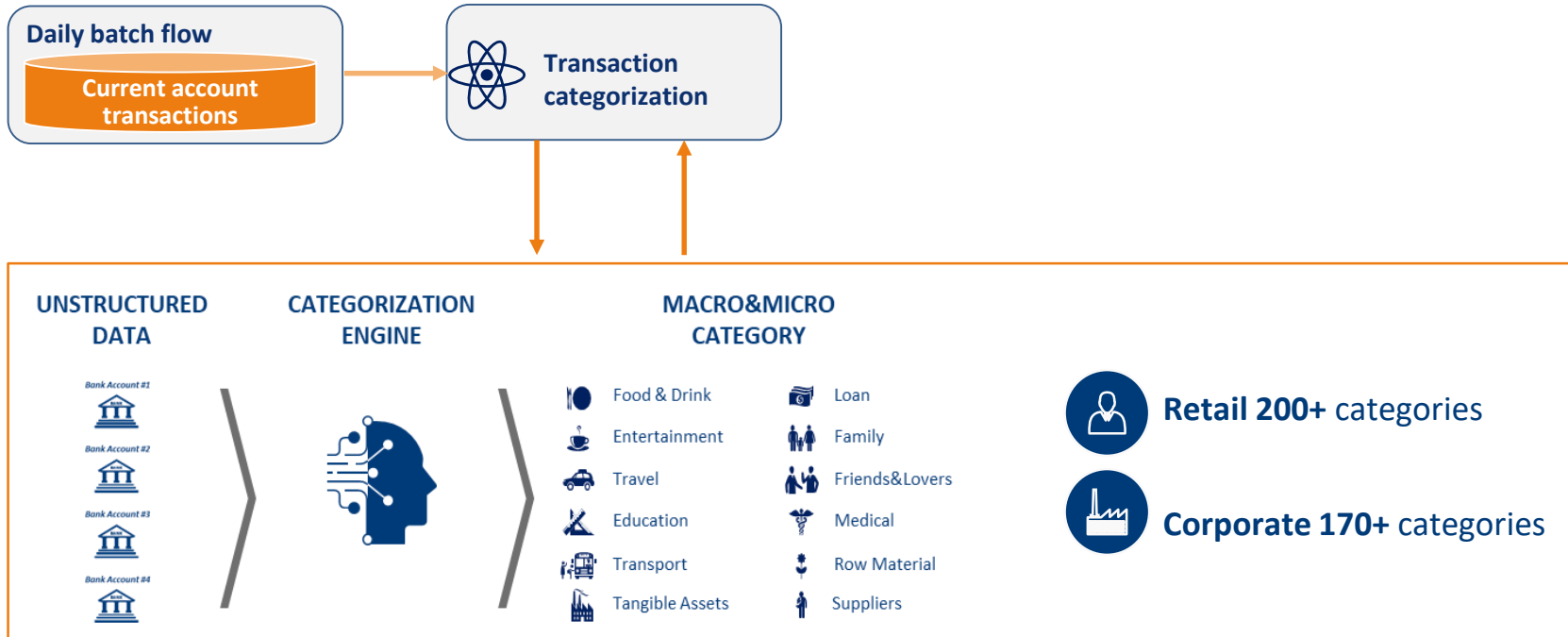
- Up to **15% risky exposure** preemptively managed
- Up to **-10% cost of risk**



# EARLY WARNING SYSTEM WITH TRANSACTION DATA

AI driven categorization and ML modeling of categorized data

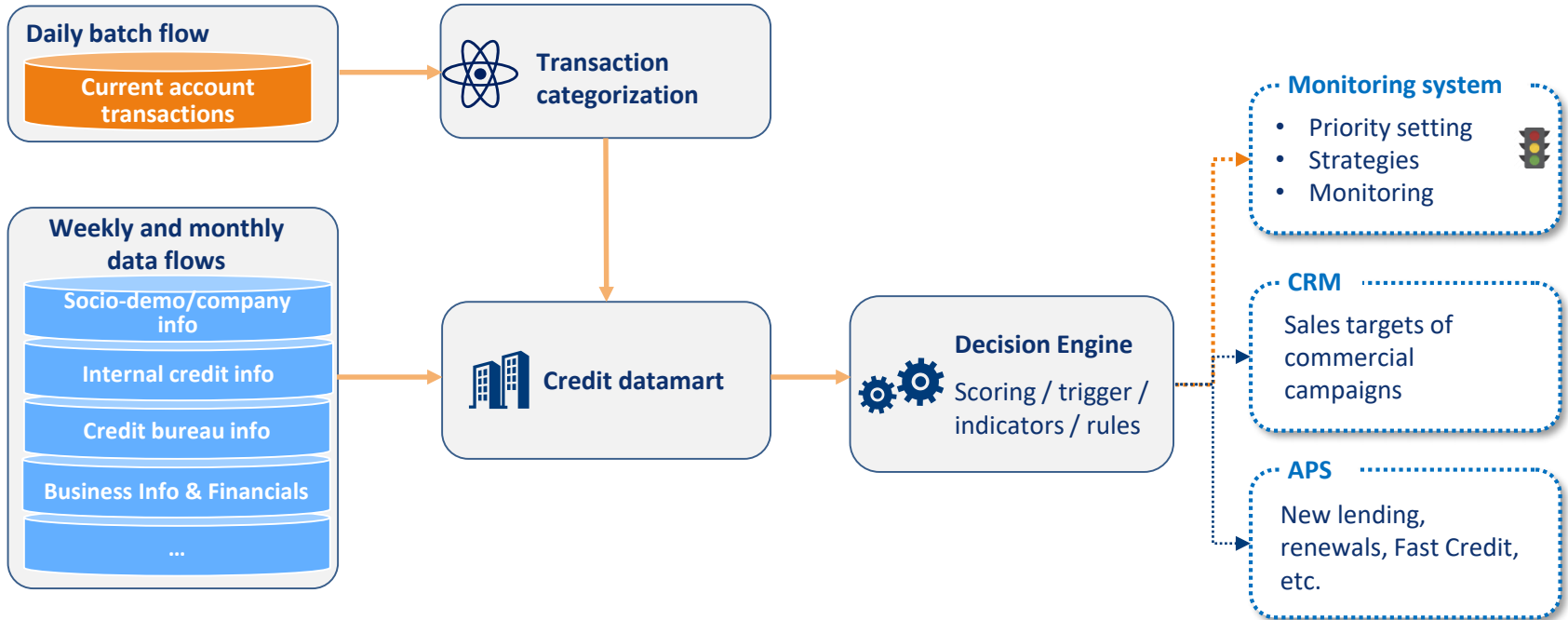
*Simplified flow, illustrative*



# EARLY WARNING SYSTEM WITH TRANSACTION DATA

## AI driven categorization and ML modeling of categorized data

*Simplified flow, illustrative*



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**CASE 2 – SECOND LEVEL CONTROL SYSTEM WITH GEN-AI**

WRAP UP

# SECOND LEVEL CONTROL SYSTEM WITH GEN-AI

## Tier 1 banking group



### BACKGROUND

- > The Bank's Risk Management wants to evolve the current solution in order to allow for a greater analysis of loan application files still keeping constant the number of FTEs involved in the process.
- > The current process is fully manual and covers only a minor part of application files randomly selected

### SOLUTION



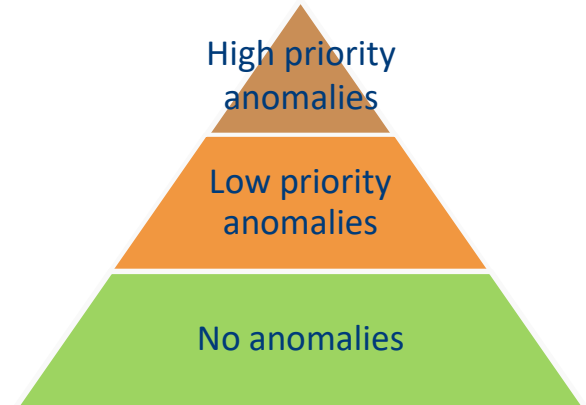
- > **Gen-AI solution** to analyze systematically all of the notes reported in loan applications, identify potential anomalies, rank them by priority grades and provide analysts with pre-assessed cases

### GOAL



- > **Increase effectiveness** of internal control systems
- > **Reduce the *psychological effect*** of analysts' assessment
- > **Structured set of anomalies**

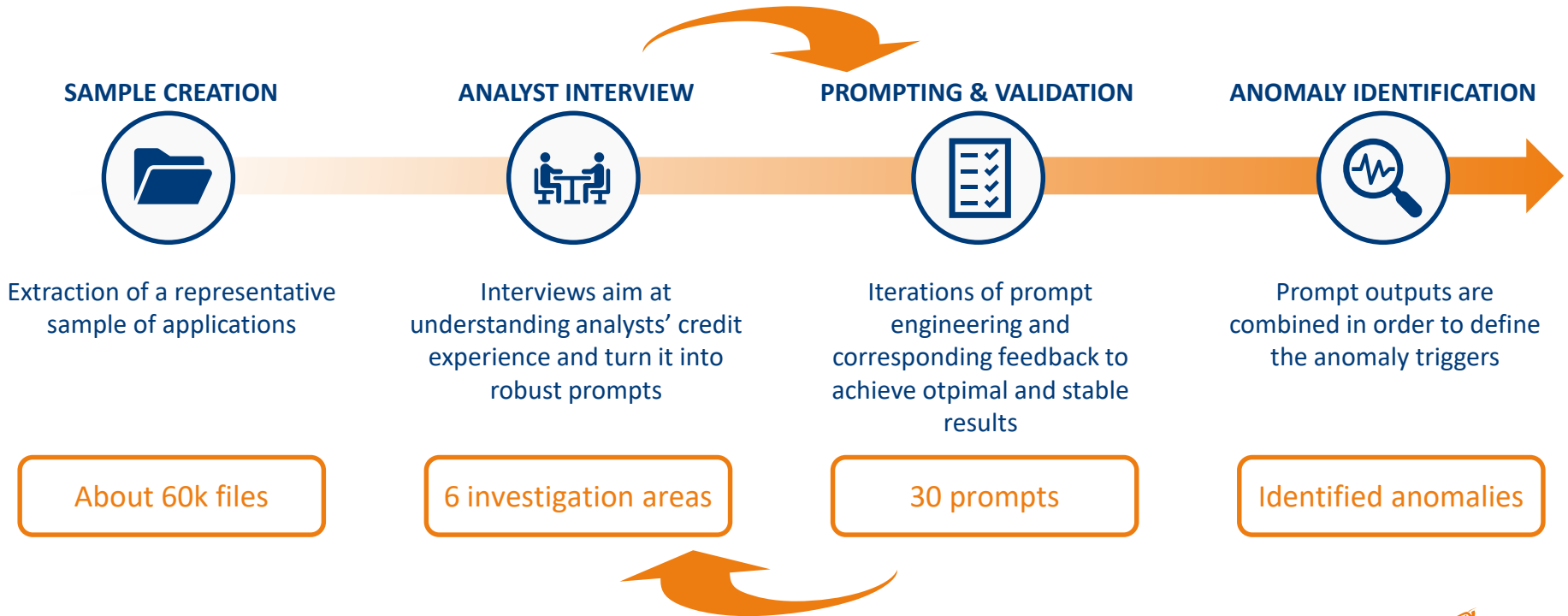
Illustrative example





# IMPLEMENTING THE GEN-AI DRIVEN CONTROLS

## The high level process overview



# DIRECTING GEN-AI CONTROLS

## Definition of the investigation areas: a 2-phase approach



### Phase 1

- 1 Credit policies**  
Bank's strategy towards the customer (e.g. Develop, reduce, etc.) and coherence of analysts' comments
- 2 UTP**  
NPL/UTP exposures detected by the analyst
- 3 Cost structure**  
Issues with high costs, with tax duties, and operating inefficiencies

- 4 Cash flows**  
Analysis of cashflows with focus on debt sustainability and future revenues
- 5 Conditional decision**  
Identification of conditions set by the analysts and to which decision is subject
- 6 Legal proceedings**  
Identification of past and current legal proceedings identified by the analyst

### Phase 2

- 7. Collateral assessment**
- 8. Inventories**
- 9. ...**
- 10. ...**
- 11. ...**

# ANALYSIS OF INVESTIGATION AREAS WITH GEN- AI

A real example – Reading, understanding and managing comments

## Application 1

### Proposed Note:

The customer asks us for an increase in the cash credit limit of XXXX euros .... to cope with the expenses for the goods warehouse in anticipation of the summer season .... customers are reported for X unpaid installments on the financing of the remaining XXXXX euros. Heard, they told us that it is a dispute with XXXXXXXX, which financed the ..... supply of the company XXXXX with which customers had signed a supply contract with a fixed monthly rate: however, this company XXXXXXXX would have stopped providing the service since ..... As a result, customers have had to sign a new ..... supply contract with XXXXXXXX, while XXXXXXXX continues to require payment of the monthly installment for the service not provided by XXXXXXXX

### Note Resolution:

AGREES, IN SPITE OF THE NEGATIVITIES ARGUED IN THE PROPOSAL NOTES AND THE RESULTS OF CREDIT POLICIES, ON THE BASIS OF THE EXPERIENCED KNOWLEDGE OF THE COUNTERPARTY.

## Flag Extraction

- ✕ CREDIT POLICIES
- ✗ UTP
- ✕ COST STRUCTURE
- ✕ CASH FLOW
- ✕ CONDITIONAL DECISION
- ✗ LEGAL DISPUTES

## Anomaly Triggers

**HIGH**  
PRIORITY

**MEDIUM**  
PRIORITY

**LOW**  
PRIORITY

## Action

...

# ANALYSIS OF INVESTIGATION AREAS WITH GEN- AI, RESULTS

Different anomalies and their combinations can be found, triggering different priorities

*Illustrative*

	UTP	COST STRUCTURE	CASH FLOWS	CONDITIONAL DECISION	LEGAL PROCEEDINGS	TOTAL
HIGH						2%
MEDIUM	/				/	8%
LOW	/	/	/		/	15%
NO ANOMALIES	/	/	/	/	/	75%

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# SOME FINAL THOUGHTS



- “**Technology is neither good nor bad; nor is it neutral**” (Kranzberg’s Law #1<sup>1</sup>)
- Industry and regulators are all aware of the benefits that technology may make available and the risks linked to it
- However, this is an **irreversible process**: it is estimated that Amazon, Microsoft, Meta and Google will have invested more than 200 billions dollars by end of 2024. That is a clear sign of where we are going
- “**Technology is a very human activity – and so is the history of technology** ” (Kranzberg’s Law #6)

<sup>1</sup> Kranzberg, M. (1986). Technology and History: “Kranzberg’s Laws.” *Technology and Culture*, 27(3), 544–560

For further information



THANK YOU!



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