Using Computer Simulation for Research in Financial Fraud

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Agenda

Introduction

Simulating Financial Transactions

Case studies

Conclusions



- Introduction
 - Financial Fraud

Financial Fraud Detection

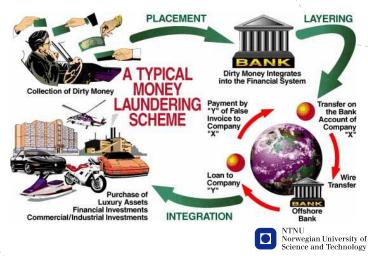
- Financial fraud can be defined as an intentional act of deception involving financial transactions for personal gain.
- The result of this is a financial loss for a corporation or a person.
- Many financial institutions have implemented controls to prevent fraud.
- But most of these controls are abused by criminals
- When prevention fails, the victim is responsible for reporting the fraud to the financial institution.
- Money Laundering is a particularly complex case of financial fraud and it consists of disguising the proceeds of criminal activities, to make them appear as if they originated from a legitimate source Norwegian University of

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Money Laundering



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The problem of applying effective controls

- By law, financial institutions protect the financial information of their customers; but at the same time they need to control and report suspicious or fraudulent behaviour.
- If you are not working inside a financial institution, then it is difficult to obtain financial datasets for testing these controls.
- Financial Institutions have internal policies to protect customer data, even from their own employees.
- Even inside a financial organization, it is difficult to develop effective controls without going through many cycles of trial and error with business operations.
- Finally, researchers in the specific field of financial fraud encounter many during their research.



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- Introduction
 - Financial Fraud

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No Data ... What to do now?

- Several years ago, we started to address some of the problems in developing novel methods for fraud detection in general and for fraud detection for Mobile Money Payments in particular.
- Unfortunately, this Mobile Money Service was not producing any data to further our research.
- Instead of waiting for the data to come, we started to develop a data generation method that is based on simulation of financial transactions.
- We worked within the domains of retail sales and financial services.
- We used computer simulations for research in the area of financial services.



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- Introduction
 - ∟_{Financial} Fraud
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I have a tool for you



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- -Introduction
 - $__{\text{Simulation}}$
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Simulator ... Simulation

- Computer Simulation is used in many different fields and domains to infer conclusions about the behaviour of real-worldphenomena.
- For example: weather prediction, logistics, thermodynamics, electronics, etc.
- It is a relatively new development for fraud detection in financial services,
- mainly due to the privacy issues explained earlier.
- Researchers had little access to data, while law enforcement authorities had other concerns (preventing crime).



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- Introduction
 - ∟_{Simulation}

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Multi-Agent Based Simulation (MABS)

- Multi-Agent Based Simulation (MABS) is a specific type of Computer Simulation
- MABS are built from the bottom up.
- The design doesn't need to know the complex structural behaviour or emerging behaviour that it is simulating.
- It uses knowledge of individual behavior and then creates collectives of these individuals: these collectives exhibit complex emerging behaviour.
- This makes it very useful to model financial services behaviour that emerges from these collections of individual customers; this emergence produces complex financial interactions.
 - MABS keywords: Agents, Environment, States, Behaviour,

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Steps

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- Introduction
 - └_synthetic data

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Why Synthetic Data?

There are many benefits of using synthetic datasets:

- Data is ready and available
- Privacy of customers is not affected
- Results can be disclosed to, and compared by, other researchers
- Different scenarios can be modeled using well controlled parameters.
- We can avoid the class imbalance problem for ML classification.
- Prior knowledge of the interesting (fraud) cases avoids the problem of mislabeled classes for ML.



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- Introduction
 - └_synthetic data

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Why Synthetic Data?

Unfortunately, there are still issues that arise when using synthetic datasets:

- Data generated might not be representative or realistic
- Data can be biased
- It is difficult to build a fully realistic model, due to the complexity of the variables and parameters.
- Any fraud detected can be used only as an example and does not represent a real case of fraud.
- The methods discovered using synthetic datasets still need to be retested with real world datasets.



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A Financial Simulator is our Tool

- We realized that simulation was needed for fraud research into financial transactions.
- Like an astronomer, we started to build our own telescope to see the stars.
- We built 3 simulators that worked as our "telescope" for research in financial transactions: RetSim, BankSim and PaySim.
- We developed a method to take advantage of our tool to produce synthetic datasets and to address most of the issues that arise with it.



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Simulating Financial Transactions

As a summary our method follows these steps in order to simulate financial transactions and perform experiments:

- 1. Obtain a sample of real data.
- 2. Perform data analysis to extract aggregated information for input into the simulator.
- 3. Add parametrization for expected fraud scenarios.
- 4. Run the simulator several times, using different random number seeds and/or different fraud configurations.
- 5. Apply the fraud detection methods on the generated synthetic dataset.
- 6. Summarize the results and performance from the experiments.
- 7. Repeat from step 3 on, for various fraud scenarios.

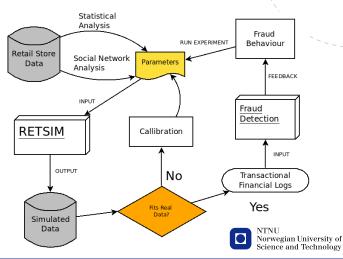


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Simulating Financial Transactions

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Case Study: RetSim



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- 1. Obtain a sample of real data
 - This is probably the hardest part of the whole method
 - Real data is important, because without it the simulationresults can not be trusted.
 - Fraud detection results are highly dependent on the dataset.
 - For better results, a sample dataset that represents the financial service is required
 - This means that it covers enough (interesting) periods of time to learn more during the data analysis.
 - If there is fraud, it should be properly labeled and identified with respect to which class of fraud it belongs in.



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- 1. Obtain a sample of real data
 - The real data sample can be obtained in several different ways:
 - Full dump of a database (100% access)
 - All of the data over a period of time
 - Partial attributes of the data for some period of time
 - All data anonymised with respect to customer information
 - Anonymisation by adding noise corruption (lowers the data quality)
 - Simply aggregating information over a period of time



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2. Perform a data analysis to extract aggregated information for input to the simulator.

- Depending on the way the real data is provided, we need to perform several operations to convert the data into the format required
- The simulator uses aggregations of information over a period of time, as input.
- The time granularity of the aggregation is specified on the simulation as a STEP
- To accurately mimic the data distribution, we must extract aggregated information from the original data that matches each step in the simulation
- There are also initial values and other input values extracted from the real data

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Simulating Financial Transactions

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2. Perform a data analysis to extract aggregated information for simulator input.

- The information extracted is represented in terms of probabilities to ease the decision processes of the agents.
- Social Network Analysis (SNA) helps to recreate the topology of the customers' relations inside the simulation.
- The agent interacts with other agents within the environment and this interaction is specified by the information extracted during the SNA
- The data analysis can also be done by employees of the financial institution that have access to the sample
- Researchers only need the output of this step, to continue the process, this allows financial institutions to preserve the privacy of the customers

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- 3. Add parametrization about expected fraud scenarios
 - The simulators are usually built to serve a purpose.
 - Our simulators contain agents that, under certain conditions, act contrary to the law.
 - The synthetic dataset has the benefit that can be generated (according to the researcher's needs to study how certain fraud, might affect a specific scenario.
 - It can be a representation of the original dataset (sample).
 - That is why we extract the aggregated information from the sample.
 - Part of the simulator validation is to show that, given certain parameters, we can reproduce similar datasets.



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4. Run the simulator several times using different random number seeds and/or different fraud configurations.

- In order to perform research in this field we need to be able to test different configurations
- The Financial Simulator can also be used to answer all the "WHAT IF" questions that are common during research
- Researchers can run the simulator several times, using controlled variation on the parameters, to create new scenarios with normal and fraudulent data.
- This is specifically useful for answering questions such as WHAT IF: There is no fraud, there is little fraud, there is a lot of fraud, double the number of customers, and so on.



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- 5. Apply the fraud detection methods on the generated synthetic dataset
 - This is one of the most important steps in the method.
 - By changing the parameters in the previews step, we can generate diverse scenarios
 - These scenarios produce datasets with data that are labeled as fraudulent or not fraudulent (as appropriate).
 - Once a dataset is generated, different methods for fraud detection can be tested and evaluated using the fraud label.
 - A method for fraud detection can also be tested and evaluated with different scenarios that use the same fraud label.
 - Fraud prevention methods can be also be added to the simulator to test and evaluate against fraud scenarios with the same flagged fraud.



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Simulating Financial Transactions

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6. Summarizing the results and performance from the experiments

- The biggest advantage of using a simulator over a real dataset is that we know with certainty how much fraud is present and where it is located.
- In a real dataset, it is impossible to guaranty that there isn't any undetected hidden fraud.
- Since we control our malicious agents, we can flag *all* fraudulent behaviour, because we have prior knowledge about the level of fraud injected into the dataset.
- Measuring all the fraud present in a dataset is one of the biggest challenges when using real data, but not with synthetic data.



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- 7. Repeat from step 3 for different fraud scenarios
 - After analysing the results on the previous step new questions may arise
 - New scenarios can be generated
 - Fraud detection methods can be modified to improve the results
 - A simulator can be used in a loop to improve results and perform research in fraud detection
 - Re-Starting at step 3 is more effective for research but some researchers might chose to work on a previously generated dataset (step 5) to test different methods and compare results against previous research.



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- Case studies
 - L_{RetSim}

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The RetSim Simulator

- RetSim (Retail Store Simulator) is an agent-based simulator of a shoe store
- It is based on transactional data from one of the largest retail shoe sellers in Sweden
- We received access to a full dump of the dataset
- During the data analysis step, we selected a store that contained sufficient information about customers to enable performance of SNA to build a network
- One year of transactions was selected as the period of time for analysis.
- Each step on the simulation represents a day of sales on the store.

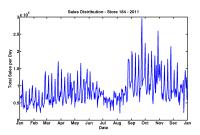


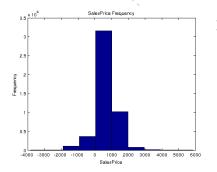
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- $_$ Case studies
 - L_{RetSim}

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Data Analysis Sales





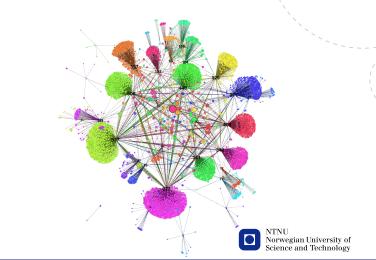


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- └─Case studies
 - ${}_{\rm RetSim}$

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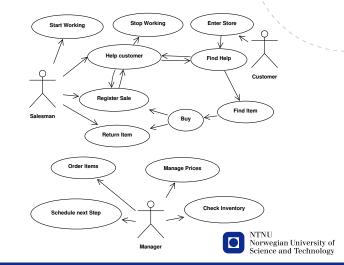
Data Analysis SNA



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Model



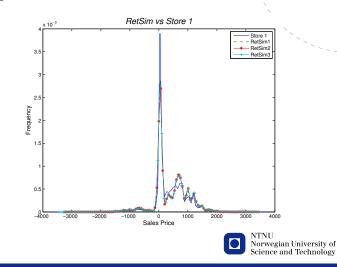
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- $\Box_{\text{Case studies}}$
 - ${}_{\rm RetSim}$

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Calibration



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Fraud Scenarios

We injected 2 fraud scenarios:

- The Refunds scenario includes cases where the salesman creates fraudulent refund slips, keeping the cash refund for him- or herself.
- The refund scenario was simulated by estimating the average number of refunds per sale and the corresponding standard deviation
- Coupon reductions/discounts scenario includes cases where the salesman registers a discount on the sale without telling the customer
- i.e., the customer pays the full sales price, and the salesman keeps the difference.



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Thresholds everywhere

- We have our "telescope", now its time for "astronomy".
- One of the more common techniques to detect fraud is the use of limits or thresholds.
- So we used the RetSim simulator to dig into this topic in the fraud scenarios modeled.
- We live in a world where machine learning is assisting in many fields of human endeavor.
- Fraud is one of the areas that ML is being used.
- How effective are threshold methods and what is the margin of improvement from machine learning methods?
- Do ML methods have improved cost over threshold methods?

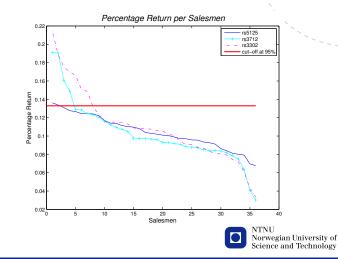


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- Case studies
 - ${}_{\rm RetSim}$

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Using Thresholds



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RetSim gets more useful ideas

- Why not use it for measuring the cost of fraud?
- Fraud cost is usually estimated by using statistics.
- But it's not very accurate and it cannot predict future scenarios
- We use our "telescope" to explore unseen areas of the "universe".
- This idea of "the cost of fraud" will motivate managers to invest in security

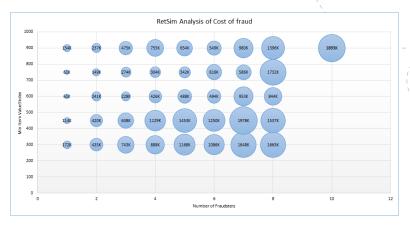


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- └─Case studies
 - ∟_{RetSim}

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Using RetSim to measure the cost of fraud



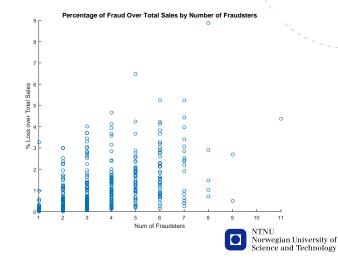


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- └─Case studies
 - L_{RetSim}

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Using RetSim to measure the cost of fraud



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- Case studies
 - BankSim

The BankSim Simulator

- BankSim is a Bank Payment Simulation for Fraud Detection Research
- BBVA bank in Spain was sharing a webservice to query aggregated financial transactions with the purpose of developing apps for a contest.
- We participated in this contest
- But the first thing we did, was to query all available data and store it locally for our research
- Then we built "The BankSim Simulator" that used the aggregated information
- The bank never disclosed any information about their customers when it would have been useful for us to build a consumption model.



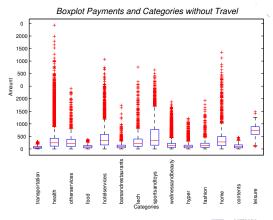
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└─Case studies

∟_{BankSim}

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Banksim simulated categories





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- Case studies
 - ∟_{PaySim}

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PaySim: Financial Simulator of Mobile Money

- PaySim is a Financial Simulator of Mobile Money for Fraud Detection Research
- This was our most recent simulator; it was built just last year.
- We obtained a data sample with more than 24 million transactions from over a period of 1 month.
- We covered 5 of the most important transaction types: CASH-IN, CASH-OUT, DEBIT, PAYMENT and TRANSFER.
- We followed the distributions and simulated the normal behaviour
- We later decided to study the case of fraud that occurs when a customer loses control over their account



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└─Case studies └─_{PavSim}

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PaySim: Financial Simulator of Mobile Money

- For the criminal to empty the account, he needs to use a merchant to cash out the account
- If there are controls to avoid daily withdraws higher than a threshold, the criminal needs to use several mule accounts.
- We simulated 4 scenarios with different thresholds in order to investigate the effectiveness of a control that prevents the emptying of an account.
- If the accounts executes 3 withdraws for the max daily amount, then the account is locked and the customer must contact customer services to unlock the account.



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Case studies

 L_{PaySim}

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PaySim Results

Table 2	Fraud	Detection	Classification
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LogName	Class	Count	Amount	% count	% amount
PS89745 (300k)	FN	27,412	6,724M	1.005%	0.363%
	FP	982	214M	0.036%	0.012%
	TN	2,607,642	1,816,764M	95.579%	98.162%
	TP	92,211	27,076M	3.380%	1.463%
PS80775 (600k)	FN	24,400	11,291M	0.990%	0.581%
	FP	58	17M	0.002%	0.001%
	TN	2,396,684	1,907,409M	97.239%	98.126%
	TP	43,604	25,114M	1.769%	1.292%
PS00273 (900k)	FN	21,072	12,854M	1.024%	0.768%
	FP	8	1M	0.000%	0.000%
	TN	2,011,006	1,639,699M	97.712%	97.903%
	TP	26,006	22,264M	1.264%	1.329%
PS98516 (1200k)	FN	20,493	16,189M	0.921%	0.858%
	FP	1	0.168M	0.000%	0.000%
	TN	2,186,516	1,849,707M	98.215%	97.993%
	TP	19,248	21,686M	0.865%	1.149%



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└─Case studies └─PavSim

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PaySim Results

- If the thresholds are set too high, then the most of the accounts won't be locked during fraud.
- If the thresholds are set too low, then several true customers will see their account locked after normal transactions.

LogName	Precision	Recall
PS89745	98.946%	77.085%
PS80775	99.867%	64.120%
PS00273	99.969%	55.240%
PS98516	99.995%	48.434%

Table 3 Fraud Detection Results



-Conclusions

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Sharing our data at Kaggle.com



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-Conclusions

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Conclusions

- Fraud detection in financial transactions is affected by the availability of datasets for testing methods
- Our approach presents an alternative: working with synthetic datasets to allow researchers to generate data from diverse scenarios and model fraud.
- Most of our results have been presented with RetSim
- PaySim is just starting to grow and collaboration with other researchers has already been started
- We aim to increase the quality of the synthetic dataset by incorporating more detailed SNA into the parameters
- We aim to use different real datasets to enrich the availability of synthetic datasets



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└─ Conclusions

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This is IT!!!

- Any questions?



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